

**CURRICULUM FRAMEWORK AND SYLLABUS**  
**FOR**  
**FIVE YEAR INTEGRATED M.Sc. (DATA SCIENCE) DEGREE PROGRAMME**  
**IN CHOICE BASED CREDIT SYSTEM**  
**FOR THE STUDENTS ADMITTED FROM THE**  
**ACADEMIC YEAR 2023-2024 ONWARDS**



**THIAGARAJAR COLLEGE OF ENGINEERING**  
(A Government Aided Autonomous Institution affiliated to Anna University)

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**THIAGARAJAR COLLEGE OF ENGINEERING, MADURAI 625 015**

**DEPARTMENT OF APPLIED MATHEMATICS AND COMPUTATIONAL SCIENCE**

**VISION**

“Academic and research excellence in Computational Science”

**MISSION**

**As a Department, we are committed to**

- Achieve academic excellence in Computational Science through innovative teaching and learning processes.
- Enable the students to be technically competent to solve the problems faced by the industry.
- Create a platform for pursuing inter-disciplinary research among the faculty and the students to create state of art research facilities.
- Promote quality and professional ethics among the students.
- Help the students to learn entrepreneurial skills.

### **Programme Educational Objectives (PEO)**

#### **Post graduates of M.Sc.(Data Science) program will be**

- PEO1: Utilizing strong quantitative aptitude and domain knowledge to apply quantitative modeling and data analysis techniques to provide solutions to the real-world business problems.
- PEO2: Applying research and entrepreneurial skills augmented with a rich set of communication, teamwork, and leadership skills to excel in their profession.
- PEO3: Showing continuous improvement in their professional career through life-long learning, appreciating human values and ethics.

### **Programme Outcomes (PO) for M.Sc. (Data Science)**

On completion of the programme, the students are expected to

1. **Engineering knowledge:** Apply the knowledge of mathematics, science, engineering fundamentals, and an engineering specialization to the solution of complex engineering problems.
2. **Problem analysis:** Identify, formulate, review research literature, and analyze complex engineering problems reaching substantiated conclusions using first principles of mathematics, natural sciences, and engineering sciences.
3. **Design/development of solutions:** Design solutions for complex engineering problems and design system components or processes that meet the specified needs with appropriate consideration for the public health and safety, and the cultural, societal, and environmental considerations.
4. **Conduct investigations of complex problems:** Use research-based knowledge and research methods including design of experiments, analysis and interpretation of data, and synthesis of the information to provide valid conclusions.
5. **Modern tool usage:** Create, select, and apply appropriate techniques, resources, and modern engineering and IT tools including prediction and modeling to complex engineering activities with an understanding of the limitations.
6. **The engineer and society:** Apply reasoning informed by the contextual knowledge to assess societal, health, safety, legal and cultural issues and the

Five year Integrated M.Sc (Data Science) Degree Programme 2023-2024 onwards  
consequent responsibilities relevant to the professional engineering practice.

7. **Environment and sustainability:** Understand the impact of the professional engineering solutions in societal and environmental contexts, and demonstrate the knowledge of, and need for sustainable development.
8. **Ethics:** Apply ethical principles and commit to professional ethics and responsibilities and norms of the engineering practice.
9. **Individual and team work:** Function effectively as an individual, and as a member or leader in diverse teams, and in multidisciplinary settings.
10. **Communication:** Communicate effectively on complex engineering activities with the engineering community and with society at large, such as, being able to comprehend and write effective reports and design documentation, make effective presentations, and give and receive clear instructions.
11. **Project management and finance:** Demonstrate knowledge and understanding of the engineering and management principles and apply these to one's own work, as a member and leader in a team, to manage projects and in multidisciplinary environments.
12. **Life-long learning:** Recognize the need for, and have the preparation and ability to engage in independent and life-long learning in the broadest context of technological change.

### **Program Specific Outcomes (PSO) for M.Sc. Data Science**

**PSO1:** To become a skilled Data Scientist in industry, academia, or government by independently carrying out research based statistical experiments to solve analytical problems following the ethics for data science

**PSO2:** To exhibit professional and interpersonal skills in data collection and its analysis using mathematical modelling and statistical techniques, querying and reporting for business and societal requirements using state-of-the-art tools.

**PEO-PO & PSO Mapping:**

PEO	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
1														
2														
3														

## Five Year Integrated M.Sc (Data Science) Degree Programme Curriculum

### 2023-2024 Onwards CHOICE BASED CREDIT SYSTEM

#### Credit Distribution:

S. No.	Category of courses	Credits	Percentage of Credits To TotalCredits
1	Foundation Courses	50	23.70
2	Professional Core Courses- Theory	88	41.70
3	Professional Core Courses- Practical	28	13.27
4	Professional Elective Courses	15	7.11
5	Employability Enhancement Courses	30	14.22
<b>Total Credits</b>		<b>211</b>	<b>100%</b>

#### Foundation courses (FC):

Course Code	Name Of the Course	Category	No. of Hours / Week			Credits
			L	T	P	
<b>THEORY</b>						
23DS110	CALCULUS	FC	4	0	0	4
23DS120	FOUNDATIONS OF DATA SCIENCE	FC	4	0	0	4
23DS130	DIGITAL ELECTRONICS AND COMPUTER ORGANIZATION	FC	4	0	0	4
23DS140	PROBLEM SOLVING USING C PROGRAMMING	FC	4	0	0	4
23DS150	DISCRETE STRUCTURES	FC	4	0	0	4
23DS170	C PROGRAMMING LAB	FC	0	0	4	2
23DS210	THEORY OF PROBABILITY	FC	4	0	0	4
23DS230	GRAPH THEORY	FC	4	0	0	4
23DS310	PARTIAL DIFFERENTIAL EQUATIONS AND TRANSFORMS	FC	4	0	0	4
23DS320	ABSTRACT ALGEBRA	FC	4	0	0	4
23DS410	LINEAR ALGEBRA	FC	4	0	0	4
23DS510	NUMERICAL METHODS	FC	4	0	0	4
23DS520	OPTIMIZATION TECHNIQUES	FC	4	0	0	4

**Professional Core Courses (PCC):**

Course code	Name of the Course	Category	No. of Hours / Week			credits
			L	T	P	
23DS220	APPLIED STATISTICS	PCC	4	0	0	4
23DS240	OBJECT ORIENTED PROGRAMMING	PCC	4	0	0	4
23DS250	ORGANIZATIONAL THEORY AND BEHAVIOUR	PCC	3	0	0	3
23DS270	PYTHON PROGRAMMING AND APPLIED STATISTICS LAB	PCC	0	0	6	3
23DS280	OBJECT ORIENTED PROGRAMMING LAB	PCC	0	0	4	2
23DS330	DATA STRUCTURES	PCC	4	0	0	4
23DS340	DATABASE MANAGEMENT	PCC	4	0	0	4
23DS350	OPERATING SYSTEMS	PCC	3	0	0	3
23DS370	DATA STRUCTURES LAB	PCC	0	0	4	2
23DS380	RELATIONAL DATABASE LAB	PCC	0	0	4	2
23DS420	PREDICTIVE ANALYTICS	PCC	4	0	0	4
23DS430	DESIGN AND ANALYSIS OF ALGORITHMS	PCC	3	1	0	4
23DS440	ADVANCED DATA STRUCTURES	PCC	4	0	0	4
23DS450	SOFTWARE ENGINEERING	PCC	3	0	0	3
23DS470	PREDICTIVE ANALYTICS LAB	PCC	0	0	4	2
23DS480	JAVA PROGRAMMING LAB	PCC	0	0	6	3
23DS530	WEB TECHNOLOGY	PCC	4	0	0	4
23DS540	MACHINE LEARNING	PCC	4	0	0	4
23DS550	COMPUTER NETWORKS	PCC	4	0	0	4
23DS570	WEB TECHNOLOGY LAB	PCC	0	0	4	2
23DS610	DEEP LEARNING	PCC	4	0	0	4

23DS620	DATA MINING	PCC	4	0	0	4
23DS630	BIG DATA SYSTEMS	PCC	4	0	0	4
23DS640	ETHICS FOR DATA SCIENCE	PCC	3	0	0	3
23DS670	DEEP LEARNING LAB	PCC	0	0	4	2
23DS680	BIG DATABASE SYSTEMS LAB	PCC	0	0	4	2
23DS810	REINFORCEMENT LEARNING	PCC	4	0	0	4
23DS820	DATA VISUALIZATION	PCC	4	0	0	4
23DS830	BUSINESS ANALYTICS	PCC	3	1	0	4
23DS870	MATHEMATICAL COMPUTING LAB	PCC	0	0	4	2
23DS880	BUSINESS ANALYTICS AND VISUALIZATION LAB	PCC	0	0	4	2
23DS910	WEB ANALYTICS	PCC	4	0	0	4
23DS920	NATURAL LANGUAGE PROCESSING	PCC	4	0	0	4
23DS930	COMPUTER VISION	PCC	3	1	0	4
23DS970	WEB ANALYTICS LAB	PCC	0	0	4	2
23DS980	NATURAL LANGUAGE PROCESSING LAB	PCC	0	0	4	2

**Professional Elective Courses (PEC):**

Course code	Name of the Course	Category	No. of Hours / Week			credits
			L	T	P	
23DSPA0	HIGH PERFORMANCE COMPUTING	PEC	3	0	0	3
23DSPB0	MOBILE APPLICATION DEVELOPMENT	PEC	3	0	0	3
23DSPC0	HEALTHCARE INFORMATICS	PEC	3	0	0	3
23DSPD0	EMBEDDED SYSTEM	PEC	3	0	0	3
23DSPE0	MARKETING ANALYTICS	PEC	3	0	0	3
23DSPF0	LARGE LANGUAGE MODELS	PEC	3	0	0	3
23DSPG0	SOFT COMPUTING	PEC	3	0	0	3
23DSPH0	MATHEMATICAL MODELING	PEC	3	0	0	3
23DSPJ0	GRAPH ALGORITHMS	PEC	3	0	0	3

23DSPK0	EXPLAINABLE ARTIFICIAL INTELLIGENCE	PEC	3	0	0	3
23DSPL0	GAME THEORY	PEC	3	0	0	3
23DSPM0	SOCIAL MEDIA ANALYTICS	PEC	3	0	0	3
23DSPN0	CLOUD COMPUTING	PEC	3	0	0	3
23DSPP0	INTERNET OF THINGS AND DATA ANALYTICS	PEC	3	0	0	3
23DSPQ0	COMPUTATIONAL FINANCE	PEC	3	0	0	3
23DSPR0	ENTERPRISE INFORMATION SYSTEM	PEC	3	0	0	3
23DSPA0	RANDOMIZED ALGORITHMS	PEC	3	0	0	3
23DSPT0	PRINCIPLES OF MANAGEMENT	PEC	3	0	0	3
23DSPU0	ACCOUNTING AND FINANCIAL MANAGEMENT	PEC	3	0	0	3
23DSPV0	WIRELESS NETWORKS	PEC	3	0	0	3
23DSPW0	NETWORK SCIENCE	PEC	3	0	0	3
23DSPX0	INFORMATION RETRIEVAL	PEC	3	0	0	3

### Employability Enhancement Courses (EEC):

Course code	Name of the Course	Category	No. of Hours / Week			credits
			L	T	P	
23DS180	PROFESSIONAL ENGLISH	EEC	0	2	2	3
23DS580	MINI PROJECT	EEC	0	0	6	3
23DSP10	PROJECT WORK I	EEC	0	0	24	12
23DSP20	PROJECT WORK II	EEC	0	0	24	12

**Thiagarajar College of Engineering, Madurai – 625 015**  
**Department of Applied Mathematics and Computational Science**  
**Five Year Integrated M.Sc (Data Science) Degree Programme [2023-2024 onwards]**  
**Scheduling of Courses**

Sem-ester	Theory					Practical		Credits
1	23DS110 CALCULUS 4:0	23DS120 FOUNDATIONS OF DATA SCIENCE 4:0	23DS130 DIGITAL ELECTRONICS AND COMPUTER ORGANIZATION 4:0	23DS140 PROBLEM SOLVING USING C PROGRAMMING 4:0	23DS150 DISCRETE STRUCTURES 4:0	23DS170 C PROGRAMMING LAB 0:2	23DS180 PROFESSIONAL ENGLISH 0:3	25
2	23DS210 THEORY OF PROBABILITY 4:0	23DS220 APPLIED STATISTICS 4:0	23DS230 GRAPH THEORY 4:0	23DS240 OBJECT ORIENTED PROGRAMMING 4:0	23DS250 ORGANIZATIONAL THEORY AND BEHAVIOR 3:0	23DS270 PYTHON PROGRAMMING AND APPLIED STATISTICS LAB 0:3	23DS280 OBJECT ORIENTED PROGRAMMING LAB 0:2	24
3	23DS310 PARTIAL DIFFERENTIAL EQUATIONS AND TRANSFORMS 4:0	23DS320 ABSTRACT ALGEBRA 4:0	23DS330 DATA STRUCTURES 4:0	23DS340 DATABASE MANAGEMENT 4:0	23DS350 OPERATING SYSTEMS 3:0	23DS370 DATA STRUCTURES LAB 0:2	23DS380 RELATIONAL DATABASE LAB 0:2	23
4	23DS410 LINEAR ALGEBRA 4:0	23DS420 PREDICTIVE ANALYTICS 4:0	23DS430 DESIGN AND ANALYSIS OF ALGORITHMS 3:1	23DS440 ADVANCED DATA STRUCTURES 4:0	23DS450 SOFTWARE ENGINEERING 3:0	23DS470 PREDICTIVE ANALYTICS LAB 0:2	23DS480 JAVA PROGRAMMING LAB 0:3	24
5	23DS510 NUMERICAL METHODS 4:0	23DS520 OPTIMIZATION TECHNIQUES 4:0	23DS530 WEB TECHNOLOGY 4:0	23DS540 MACHINE LEARNING 4:0	23DS550 COMPUTER NETWORKS 4:0	23DS570 WEB TECHNOLOGY LAB 0:2	23DS580 MINI PROJECT 0:3	25
6	23DS610 DEEP LEARNING 4:0	23DS620 DATA MINING 4:0	23DS630 BIG DATA SYSTEMS 4:0	23DS640 ETHICS FOR DATA SCIENCE 3:0	23DSPX0 PROFESSIONAL ELECTIVE I 3:0	23DS670 DEEP LEARNING LAB 0:2	23DS680 BIG DATABASE SYSTEMS LAB 0:2	22
7	23DSP10 PROJECT WORK I - 0:12							12
8	23DS810 REINFORCEMENT LEARNING 4:0	23DS820 DATA VISUALIZATION 4:0	23DS830 BUSINESS ANALYTICS 3:1	23DSPX0 PROFESSIONAL ELECTIVE-II 3:0	23DSPX0 PROFESSIONAL ELECTIVE-III 3:0	23DS870 MATHEMATICAL COMPUTING LAB 0:2	23DS880 BUSINESS ANALYTICS AND VISUALIZATION LAB 0:2	22
9	23DS910 WEB ANALYTICS 4:0	23DS920 NATURAL LANGUAGE PROCESSING 4:0	23DS930 COMPUTER VISION 3:1	23DSPX0 PROFESSIONAL ELECTIVE – IV 3:0	23DSPX0 PROFESSIONAL ELECTIVE – V 3:0	23DS970 WEB ANALYTICS LAB 0:2	23DS980 NATURAL LANGUAGE PROCESSING LAB 0:2	22
10	23DSP20 PROJECT WORK II - 0:12							12
<b>Total Credits</b>								<b>211</b>

**Thiagarajar College of Engineering, Madurai**  
**Five year Integrated M.Sc (Data Science) Degree Programme [2023-2024] onwards**  
**Choice Based Credit System**  
**Courses of Study**

**FIRST SEMESTER**

Course code	Course Title	Category	No. of Hours / Week			credits
			L	T	P	
<b>THEORY</b>						
23DS110	CALCULUS	FC	4	-	-	4
23DS120	FOUNDATIONS OF DATA SCIENCE	FC	4	-	-	4
23DS130	DIGITAL ELECTRONICS AND COMPUTER ORGANIZATION	FC	4	-	-	4
23DS140	PROBLEM SOLVING USING C PROGRAMMING	FC	4	-	-	4
23DS150	DISCRETE STRUCTURES	FC	4	-	-	4
<b>PRACTICAL</b>						
23DS170	C PROGRAMMING LAB	FC	-	-	4	2
23DS180	PROFESSIONAL ENGLISH	EEC	-	2	2	3
TOTAL			20	2	6	25

**SECOND SEMESTER**

Course code	Course Title	Category	No. of Hours / Week			credits
			L	T	P	
<b>THEORY</b>						
23DS210	THEORY OF PROBABILITY	FC	4	-	-	4
23DS220	APPLIED STATISTICS	PCC	4	-	-	4
23DS230	GRAPH THEORY	FC	4	-	-	4
23DS240	OBJECT ORIENTED PROGRAMMING	PCC	4	-	-	4
23DS250	ORGANIZATIONAL THEORY AND BEHAVIOUR	PCC	3	-	-	3
<b>PRACTICAL</b>						
23DS270	PYTHON PROGRAMMING AND APPLIED STATISTICS LAB	PCC	-	-	6	3
23DS280	OBJECT ORIENTED PROGRAMMING LAB	PCC	-	-	4	2
TOTAL			19	-	10	24

**THIRD SEMESTER**

Course code	Course Title	Category	No. of Hours / Week			credits
			L	T	P	
<b>THEORY</b>						
23DS310	PARTIAL DIFFERENTIAL EQUATIONS AND TRANSFORMS	FC	4	-	-	4
23DS320	ABSTRACT ALGEBRA	FC	4	-	-	4
23DS330	DATA STRUCTURES	PCC	4	-	-	4
23DS340	DATABASE MANAGEMENT	PCC	4	-	-	4
23DS350	OPERATING SYSTEMS	PCC	3	-	-	3
<b>PRACTICAL</b>						
23DS370	DATA STRUCTURES LAB	PCC	-	-	4	2
23DS380	RELATIONAL DATABASE LAB	PCC	-	-	4	2
<b>TOTAL</b>			19	-	8	23

**FOURTH SEMESTER**

Course code	Course Title	Category	No. of Hours / Week			credits
			L	T	P	
<b>THEORY</b>						
23DS410	LINEAR ALGEBRA	FC	4	-	-	4
23DS420	PREDICTIVE ANALYTICS	PCC	4	-	-	4
23DS430	DESIGN AND ANALYSIS OF ALGORITHMS	PCC	3	1	-	4
23DS440	ADVANCED DATA STRUCTURES	PCC	4	-	-	4
23DS450	SOFTWARE ENGINEERING	PCC	3	-	-	3
<b>PRACTICAL</b>						
23DS470	PREDICTIVE ANALYTICS LAB	PCC	-	-	4	2
23DS480	JAVA PROGRAMMING LAB	PCC	-	-	6	3
<b>TOTAL</b>			18	1	10	24

**FIFTH SEMESTER**

Course code	Course Title	Category	No. of Hours / Week			credits
			L	T	P	
<b>THEORY</b>						
23DS510	NUMERICAL METHODS	FC	4	-	-	4
23DS520	OPTIMIZATION TECHNIQUES	FC	4	-	-	4
23DS530	WEB TECHNOLOGY	PCC	4	-	-	4
23DS540	MACHINE LEARNING	PCC	4	-	-	4
23DS550	COMPUTER NETWORKS	PCC	4	-	-	4
<b>PRACTICAL</b>						
23DS570	WEB TECHNOLOGY LAB	PC	-	-	4	2
23DS580	MINI PROJECT	EEC	-	-	6	3
<b>TOTAL</b>			20	-	10	25

**SIXTH SEMESTER**

Course code	Course Title	Category	No. of Hours / Week			credits
			L	T	P	
<b>THEORY</b>						
23DS610	DEEP LEARNING	PCC	4	-	-	4
23DS620	DATA MINING	PCC	4	-	-	4
23DS630	BIG DATA SYSTEMS	PCC	4	-	-	4
23DS640	ETHICS FOR DATA SCIENCE	PCC	3	-	-	3
23DSPX0	PROFESSIONAL ELECTIVE I	PEC	3	-	-	3
<b>PRACTICAL</b>						
23DS670	DEEP LEARNING LAB	PCC	-	-	4	2
23DS680	BIG DATABASE SYSTEMSLAB	PCC	-	-	4	2
TOTAL			18	-	8	22

**SEVENTH SEMESTER**

Course code	Course Title	Category	No. of Hours / Week			credits
			L	T	P	
<b>PRACTICAL</b>						
23DSP10	PROJECT WORK – I	EEC	-	-	24	12
TOTAL			-	-	24	12

**EIGHTH SEMESTER**

Course code	Course Title	Category	No. of Hours / Week			credits
			L	T	P	
<b>THEORY</b>						
23DS810	REINFORCEMENT LEARNING	PCC	4		-	4
23DS820	DATA VISUALIZATION	PCC	4	-	-	4
23DS830	BUSINESS ANALYTICS	PCC	3	1	-	4
23DSPX0	PROFESSIONAL ELECTIVE-II	PEC	3	-	-	3
23DSPX0	PROFESSIONAL ELECTIVE-III	PEC	3	-	-	3
<b>PRACTICAL</b>						
23DS870	MATHEMATICAL COMPUTING LAB	PCC	-	-	4	2
23DS880	BUSINESS ANALYTICS AND VISUALIZATION LAB	PCC	-	-	4	2
TOTAL			17	1	8	22

**NINTH SEMESTER**

Course code	Course Title	Category	No. of Hours / Week			credits
			L	T	P	
<b>THEORY</b>						
23DS910	WEB ANALYTICS	PCC	4	-	-	4
23DS920	NATURAL LANGUAGE PROCESSING	PCC	4	-	-	4
23DS930	COMPUTER VISION	PCC	3	1	-	4
23DSPX0	PROFESSIONAL ELECTIVE – IV	PEC	3	-	-	3
23DSPX0	PROFESSIONAL ELECTIVE – V	PEC	3	-	-	3
<b>PRACTICAL</b>						
23DS970	WEB ANALYTICS LAB	PCC	-	-	4	2
23DS980	NATURAL LANGUAGE PROCESSING LAB	PCC	-	-	4	2
<b>TOTAL</b>			<b>17</b>	<b>1</b>	<b>8</b>	<b>22</b>

**TENTH SEMESTER**

Course code	Course Title	Category	No. of Hours / Week			credits
			L	T	P	
<b>PRACTICAL</b>						
23DSP20	PROJECT WORK - II	EEC	-	-	24	12
<b>TOTAL</b>			<b>-</b>	<b>-</b>	<b>24</b>	<b>12</b>

**FC** : Foundation Course  
**PCC** : Professional Core Course  
**PEC** : Professional Elective Course  
**EEC** : Employability Enhancement Course

**L** : Lecture  
**T** : Tutorial  
**P** : Practical

**Note:**

- 1 Hour Lecture/week is equivalent to 1 credit
- 1 Hour Tutorial/week is equivalent to 1 credit
- 2 Hour Practical/week is equivalent to 1 credit
- Total credits to be earned for the award of degree: 211

**Thiagarajar College of Engineering, Madurai**  
**Five year Integrated M.Sc (Data Science) Degree Programme**  
**SCHEME OF EXAMINATIONS**  
**(For Students admitted in the Academic Year 2023-2024 onwards)**

**FIRST SEMESTER**

Course code	Course Title	Duration of Terminal Exam. in Hrs.	MARKS			Minimum Marks for Pass	
			Continuous Assessment *	Terminal Exam **	Max. Marks	Terminal Exam	Total
<b>THEORY</b>							
23DS110	CALCULUS	3	40	60	100	27	50
23DS120	FOUNDATIONS OF DATA SCIENCE	3	40	60	100	27	50
23DS130	DIGITAL ELECTRONICS AND COMPUTER ORGANIZATION	3	40	60	100	27	50
23DS140	PROBLEM SOLVING USING C PROGRAMMING	3	40	60	100	27	50
23DS150	DISCRETE STRUCTURES	3	40	60	100	27	50
<b>PRACTICAL</b>							
23DS170	C PROGRAMMING LAB	3	60	40	100	18	50
23DS180	PROFESSIONAL ENGLISH	3	60	40	100	18	50

**SECOND SEMESTER**

Course code	Course Title	Duration of Terminal Exam. in Hrs.	MARKS			Minimum Marks for Pass	
			Continuous Assessment *	Terminal Exam **	Max. Marks	Terminal Exam	Total
<b>THEORY</b>							
23DS210	THEORY OF PROBABILITY	3	40	60	100	27	50
23DS220	APPLIED STATISTICS	3	40	60	100	27	50
23DS230	GRAPH THEORY	3	40	60	100	27	50
23DS240	OBJECT ORIENTED PROGRAMMING	3	40	60	100	27	50
23DS250	ORGANIZATIONAL THEORY AND BEHAVIOUR	3	40	60	100	27	50
<b>PRACTICAL</b>							
23DS270	PYTHON PROGRAMMING AND APPLIED STATISTICS LAB	3	60	40	100	18	50
23DS280	OBJECT ORIENTED PROGRAMMING LAB	3	60	40	100	18	50

**THIRD SEMESTER**

Course code	Course Title	Duration of Terminal Exam. in Hrs.	MARKS			Minimum Marks for Pass	
			Continuous Assessment *	Terminal Exam **	Max. Marks	Terminal Exam	Total
<b>THEORY</b>							
23DS310	PARTIAL DIFFERENTIAL EQUATIONS AND TRANSFORMS	3	40	60	100	27	50
23DS320	ABSTRACT ALGEBRA	3	40	60	100	27	50
23DS330	DATA STRUCTURES	3	40	60	100	27	50
23DS340	DATABASE MANAGEMENT	3	40	60	100	27	50
23DS350	OPERATING SYSTEMS	3	40	60	100	27	50
<b>PRACTICAL</b>							
23DS370	DATA STRUCTURES LAB	3	60	40	100	18	50
23DS380	RELATIONAL DATABASE LAB	3	60	40	100	18	50

**FOURTH SEMESTER**

Course code	Course Title	Duration of Terminal Exam. in Hrs.	MARKS			Minimum Marks for Pass	
			Continuous Assessment *	Terminal Exam **	Max. Marks	Terminal Exam	Total
<b>THEORY</b>							
23DS410	LINEAR ALGEBRA	3	40	60	100	27	50
23DS420	PREDICTIVE ANALYTICS	3	40	60	100	27	50
23DS430	DESIGN AND ANALYSIS OF ALGORITHMS	3	40	60	100	27	50
23DS440	ADVANCED DATA STRUCTURES	3	40	60	100	27	50
23DS450	SOFTWARE ENGINEERING	3	40	60	100	27	50
<b>PRACTICAL</b>							
23DS470	PREDICTIVE ANALYTICS LAB	3	60	40	100	18	50
23DS480	JAVA PROGRAMMING LAB	3	60	40	100	18	50

**FIFTH SEMESTER**

Course code	Course Title	Duration of Terminal Exam. in Hrs.	MARKS			Minimum Marks for Pass	
			Continuous Assessment *	Terminal Exam **	Max. Marks	Terminal Exam	Total
<b>THEORY</b>							
23DS510	NUMERICAL METHODS	3	40	60	100	27	50
23DS520	OPTIMIZATION TECHNIQUES	3	40	60	100	27	50
23DS530	WEB TECHNOLOGY	3	40	60	100	27	50
23DS540	MACHINE LEARNING	3	40	60	100	27	50
23DS550	COMPUTER NETWORKS	3	40	60	100	27	50
<b>PRACTICAL</b>							
23DS570	WEB TECHNOLOGY LAB	3	60	40	100	18	50
23DS580	MINI PROJECT	-	50	50	100	25	50

**SIXTH SEMESTER**

Course code	Course Title	Duration of Terminal Exam. in Hrs.	MARKS			Minimum Marks for Pass	
			Continuous Assessment *	Terminal Exam **	Max. Marks	Terminal Exam	Total
<b>THEORY</b>							
23DS610	DEEP LEARNING	3	40	60	100	27	50
23DS620	DATA MINING	3	40	60	100	27	50
23DS630	BIG DATA SYSTEMS	3	40	60	100	27	50
23DS640	ETHICS FOR DATA SCIENCE	3	40	60	100	27	50
23DSPX0	PROFESSIONAL ELECTIVE I	3	40	60	100	27	50
<b>PRACTICAL</b>							
23DS670	DEEP LEARNING LAB	3	60	40	100	18	50
23DS680	BIG DATABASE SYSTEMS LAB	3	60	40	100	18	50

**SEVENTH SEMESTER**

Course code	Course Title	Duration of Terminal Exam. in Hrs.	MARKS			Minimum Marks for Pass	
			Continuous Assessment *	Terminal Exam **	Max. Marks	Terminal Exam	Total
<b>PROJECT WORK I</b>							
23DSP10	PROJECT WORK -I	-	150	150	300	75	150

**EIGHTH SEMESTER**

Course code	Course Title	Duration of Terminal Exam. in Hrs.	MARKS			Minimum Marks for Pass	
			Continuous Assessment *	Terminal Exam **	Max. Marks	Terminal Exam	Total
<b>THEORY</b>							
23DS810	REINFORCEMENT LEARNING	3	40	60	100	27	50
23DS820	DATA VISUALIZATION	3	40	60	100	27	50
23DS830	BUSINESS ANALYTICS	3	40	60	100	27	50
23DSPX0	PROFESSIONAL ELECTIVE-II	3	40	60	100	27	50
23DSPX0	PROFESSIONAL ELECTIVE-III	3	40	60	100	27	50
<b>PRACTICAL</b>							
23DS870	MATHEMATICAL COMPUTING LAB	3	60	40	100	18	50
23DS880	BUSINESS ANALYTICS AND VISUALIZATION LAB	3	60	40	100	18	50

**NINTH SEMESTER**

Course code	Course Title	Duration of Terminal Exam. in Hrs.	MARKS			Minimum Marks for Pass	
			Continuous Assessment *	Terminal Exam **	Max. Marks	Terminal Exam	Total
<b>THEORY</b>							
23DS910	WEB ANALYTICS	3	40	60	100	27	50
23DS920	NATURAL LANGUAGE PROCESSING	3	40	60	100	27	50
23DS930	COMPUTER VISION	3	40	60	100	27	50
23DSPX0	PROFESSIONAL ELECTIVE – IV	3	40	60	100	27	50
23DSPX0	PROFESSIONAL ELECTIVE – V	3	40	60	100	27	50
<b>PRACTICAL</b>							
23DS970	WEB ANALYTICS LAB	3	60	40	100	18	50
23DS980	NATURAL LANGUAGE PROCESSING LAB	3	60	40	100	18	50

**TENTH SEMESTER**

Course code	Course Title	Duration of Terminal Exam. in Hrs.	MARKS			Minimum Marks for Pass	
			Continuous Assessment *	Terminal Exam **	Max. Marks	Terminal Exam	Total
<b>PROJECT WORK II</b>							
23DSP20	PROJECT WORK -II	-	150	150	300	75	150

\* Continuous Assessment (CA) evaluation pattern will differ from subject to subject and for different tests. This will have to be declared in advance to students. The Department will put a process in place to ensure that the actual test paper follows the declared pattern.

\*\* Terminal Examination for theory courses will be conducted for a maximum mark of 100 and subsequently be reduced to 60 marks for the award of terminal examination marks.

\*\* Terminal Examination for laboratory courses will be conducted for a maximum mark of 100 and subsequently be reduced to 40 marks for the award of terminal examination marks

\*\* Terminal viva voce for Mini Project will be conducted for a maximum mark of 100 and subsequently be reduced to 50 marks for the award of terminal examination marks

\*\* Terminal viva voce for Project work I and II will be conducted for a maximum mark of 300 and subsequently be reduced to 150 marks for the award of terminal examination marks

**23DS110****CALCULUS**

Category	L	T	P	Credits
FC	4	0	0	4

**Preamble**

This course will enable students to analyse the behaviour of functions of single and multi-variables to solve complex engineering problems, applications of multiple integrals in various real-life problems. Also, the students will learn the nature of the sequence and series in the technical problems.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Explain the concept of Differentiation and Integration	Understand
CO2	Apply the concepts of maxima and minima for a given function with several variables, through finding stationary points and Lagrangian multiplier method for finding Maxima and Minima of a constrained problem.	Apply
CO3	Apply the suitable method to solve second and higher order Differential equations.	Apply
CO4	Apply integration concept and double integral over general Areas	Apply
CO5	Apply integration concept and triple integral over general volumes to find mass and moments.	Apply
CO6	Apply sequences and Series in the problems involving Science and Engineering with the knowledge of convergence and divergence of series using different tests.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	L	L	L	M	-	-	-	-	-	-	-	L	L	M
CO2	S	S	M	M	-	-	-	-	-	-	-	M	L	S
CO3	S	S	M	M	L	-	-	-	-	-	-	M	L	S
CO4	S	S	M	S	L	-	-	-	-	-	-	M	L	S
CO5	S	S	M	S	L	-	-	-	-	-	-	M	L	S
CO6	S	S	L	L	-	-	-	-	-	-	-	L	L	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	5	20																							2	10	-			
CO2	5	10	30						50																2	-	20			
CO3			30						50																2	5	10			
CO4													5	10	20					30				2	-	10				
CO5													5	10	20					40				-	-	20				
CO6														10	20					30				2	5	10				
Total (in %)	100						100						100						100						100					

## Syllabus

**LIMITS AND CONTINUITY:** Differentiation - Function of single variable – Limit, continuity, periodic. Integration - Absolutely integral, fundamental theorem of calculus.

**FUNCTIONS OF TWO VARIABLES:** Partial derivative, total derivatives. Taylor series about a point. Stationary points – maxima, minima and saddle points. Constrained maxima and minima – Lagrange’s multipliers method.

**ORDINARY DIFFERENTIAL EQUATIONS:** Linear Differential Equations of first order - Exact differential equations, Integrating factors, Bernoulli equations. Linear Differential Equations of second and higher order with constant coefficients and variable coefficients – Euler’s and Legendre’s type. Simultaneous equations with constant Co-efficient, Method of variation of parameters.

**MULTIPLE INTEGRALS:** Double integrals in Cartesian form - Change of order of integration – Double integrals in polar form. Triple integrals in Cartesian form, Change of variable in triple integrals into polar, spherical and cylindrical co-ordinates. Applications of multiple integrals.

**SEQUENCES AND SERIES:** Infinite Sequences – convergence, divergence. Infinite Series – convergence and divergence, comparison test, integral test, ratio test, root test. Alternating series – alternating series test, absolute and conditional convergence, Power series.

## Reference Books and Web Resources:

1. Hass M. D. J., Giordano Weir F.R., “Thomas Calculus”, Pearson Education, 14<sup>th</sup> edition, 2018.
2. Erwin Kreyszig, “Advanced Engineering Mathematics”, John Wiley & Sons, 2014.
3. Lian, Hungerford, and Holcomb “Mathematics with Applications”, Addison Wesley, 2010.
4. Michael D. Greenberg, “Advanced Engineering Mathematics”, Pearson Education, 2013.
5. <http://calculus.org>
6. <https://www.khanacademy.org/math/calculus-1>
7. <https://dlmf.nist.gov>
8. <https://www.siam.org/>

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>Limits and Continuity</b>	
1.1	Differentiation - Function of single variable	2
1.2	Limit, continuity, periodic functions.	2
1.3	Integration – Absolutely integral, fundamental theorem of calculus.	2
<b>2</b>	<b>Functions of Two Variables</b>	
2.1	Partial derivative, total derivatives, Taylor series about a point.	2

Module No.	Topic	No. of Periods
2.2	Stationary points – maxima, minima and saddle points.	2
2.3	Constrained maxima and minima – Lagrange’s multipliers method.	2
<b>3</b>	<b>Ordinary Differential Equations</b>	
3.1	Linear Differential Equations of first order - Exact differential equations, Integrating factors, Bernoulli equations.	2
3.2	Linear Differential Equations of second and higher order with constant coefficients and variable coefficients	2
3.3	Euler’s and Legendre’s type.	2
3.4	Simultaneous equations with constant Co-efficient, Method of variation of parameters.	2
<b>4</b>	<b>Multiple Integrals: Double Integrals</b>	
4.1	Double integrals in Cartesian form – Problems.	3
4.2	Change of order of integration - Problems.	2
4.3	Double integrals in polar form – Problems.	2
4.4	Triple integrals in Cartesian form – Problems.	3
4.5	Change of variable in triple integrals into polar – Problems.	2
4.6	Spherical and cylindrical co-ordinates. Applications of multiple integrals– Problems.	3
<b>5</b>	<b>Sequences and Series</b>	
5.1	Infinite Sequences – convergence, divergence.	3
5.2	Infinite Series – convergence and divergence, comparison test, integral test, ratio test, root test.	3
5.3	Alternating series – alternating series test, absolute and conditional convergence, and Power series.	3
	Total	44

**Course Designer(s):**

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**23DS120 FOUNDATIONS OF DATA SCIENCE** Category L T P Credit  
FC 4 0 0 4

**Preamble**

Data science is an interdisciplinary field that draws on skills from mathematics, computer science, and statistics. This course will enable students to learn the fundamental concepts circumventing data science, and its applications.

**Prerequisite**

NIL

**Course Outcomes**

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

CO1	Describe the taxonomy of Data	Understand
CO2	Explore the current practices in Data Analytics	Understand
CO3	Identify the key roles for the Data Ecosystem	Understand
CO4	Identify the key roles for a successful analytics project	Understand
CO5	Apply the Data Analytics Life Cycle components to data science projects.	Apply
CO6	Apply data preparation and modelling techniques to data science related problem specifications.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	L	-	-	-	-	-	-	-	-	-	-	M	L
CO2	M	L	-	-	-	-	-	-	-	-	-	-	M	L
CO3	M	L	-	-	-	-	-	-	-	-	-	-	M	L
CO4	M	L	M	-	-	-	-	-	-	-	-	-	M	L
CO5	S	S	M	-	-	-	-	-	-	-	-	-	M	L
CO6	S	S	L	-	-	-	-	-	-	-	-	-	M	L

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	5	20																							4	10				
CO2	10	25					50																		4	10				
CO3	15	25					50																		4	10				
CO4													10	10										4	10					
CO5													5	10	20					50				2		20				
CO6													5	10	30					50				2		20				
Total (in %)	100						100						100						100											

## Syllabus

**Taxonomy of Data:** Basics of Data Structures - Overview of Big Data –Introduction to analytics - Data Repositories.

**State of the Practice in Analytics-** BI versus Data science, Current Analytical Architecture, Drivers of Big Data, Emerging big data ecosystem and new approach to analytics.

**Key Roles for the New Big Data Ecosystem-** Deep Analytical Talent, Data Savvy Professional, Technology and Data Enablers, Data Scientist, Examples of Big Data Analytics.

**Data Analytics-** Overview - Analytics in a Data Science Project - Key roles for a successful analytics project.

**Data Analytics Life Cycle (DALC) – Overview-** Different phases in a DALC - **Phase I Discovery-** Learning the Business Domain, Resources, Framing the Problem, Identifying key stakeholders, Interviewing the Analytics Sponsor, Developing initial hypotheses, Identifying potential data Sources.

**Phase II-Data Preparation-** Preparing the analytic Sandbox, Performing ETLT, Learning about the data, Data conditioning, Survey and Visualize data using common tools for the Data Visualization Phase.

**Phase III-Model Planning-** Data exploration and variable selection, Model selection, Common tools for the model planning phase.

**Phase IV-Model Building-** Common tools for the model building phase, **Phase V-Communicate Results, Phase VI-Operationalize.**

## Case Studies on Applications of Data Science and Data Analytics

### Reference Books

1. Wiley., Dietrich, D., Heller, B., & Yang, B, “Data science & big data analytics: discovering, analyzing, visualizing and presenting data”, Wiley, First Edition, 2015.
2. Eliot P. Rezn, “Big Data: A Beginner’s Guide to Using Data Science for Business”, CreateSpace Independent Publishing Platform, 2017.
3. Fahl, J, “Data Analytics: A Practical Guide To Data Analytics For Business, Beginner To Expert” , CreateSpace Independent Publishing Platform , 2017.

### Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>Taxonomy of Data</b>	1
1.1	Basics of Data Structures	1
1.2	Overview of Big Data	1
1.3	Introduction to analytics	1
1.4	Data Repositories	1

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>2</b>	<b>State of the Practice in Analytics</b>	
2.1	BI versus Data science	1
2.2	Current Analytical Architecture	2
2.3	Drivers of Big Data	1
2.4	Emerging big data ecosystem and new approach to analytics	2
<b>3</b>	<b>Key Roles for the New Big Data Ecosystem</b>	
3.1	Deep Analytical Talent	2
3.2	Data Savvy Professional	1
3.3	Technology and Data Enablers	1
3.4	Data Scientist	1
3.5	Examples of Big Data Analytics	1
<b>4</b>	<b>Data Analytics</b>	
4.1	Overview	1
4.2	Analytics in a Data Science Project	1
4.3	Key roles for a successful analytics project	1
<b>5</b>	<b>Data Analytics Life Cycle (DALC)</b>	
5.1	Overview	1
5.2	Different phases in a DALC	1
<b>6</b>	<b>Phase I Discovery</b>	1
6.1	Resources	1
6.2	Framing the Problem	1
6.3	Identifying key stakeholders	1
6.4	Interviewing the Analytics Sponsor	1
6.5	Developing initial hypotheses	1
6.6	Identifying potential data Sources	1
<b>7</b>	<b>Phase II-Data Preparation</b>	1
7.1	Preparing the analytic Sandbox	1
7.2	Performing ETLT	1
7.3	Learning about the data, Data conditioning	2

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
7.4	Survey and Visualize data using common tools for the Data Visualization Phase.	2
<b>8</b>	<b>Phase III-Model Planning</b>	1
8.1	Data exploration and variable selection	2
8.2	Model selection	1
8.3	Common tools for the model planning phase	1
<b>9</b>	<b>Phase IV-Model Building</b>	1
9.1	Common tools for the model building phase	1
<b>10</b>	<b>Phase V-Communicate Results,</b>	2
<b>11</b>	<b>Phase VI-Operationalize.</b>	2
<b>12</b>	<b>Case Studies on Applications of Data Science and Data Analytics</b>	2
	<b>Total</b>	<b>45</b>

**Course Designer(s):**

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**23DS130 DIGITAL ELECTRONICS AND  
COMPUTER ORGANIZATION**

Category	L	T	P	Credit
FC	4	0	0	4

**Preamble**

This course will enable students to focus on the design and analysis of electronic circuits, understand the computer hardware and software components, their interactions and perform various computational tasks.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Summarize the aspects of digital electronics in the design of simple and advanced computer architectures	Understand
CO2	Design combinational and sequential logic circuits, including adders, subtractors, encoders, decoders, multiplexers, de-multiplexers, flip-flops, counters, and shift registers.	Apply
CO3	Implement number systems and codes with appropriate number representations to perform arithmetic operations	Apply
CO4	Apply fundamental concepts of digital logic design in implementing Arithmetic and Logic unit	Apply
CO5	Demonstrate the design of control unit with micro instructions	Apply
CO6	Design input/output organizations using interrupts, DMA to fetch instructions in memory	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	-	-	-	-	-	-	-	-	-	-	L	L
CO2	S	M	-	-	-	-	-	-	-	-	-	-	L	L
CO3	S	M	-	-	-	-	-	-	-	-	-	-	L	L
CO4	S	M	M	-	-	-	-	-	-	-	-	-	L	L
CO5	S	M	L	-	-	-	-	-	-	-	-	-	L	L
CO6	S	M	L	-	-	-	-	-	-	-	-	-	L	L

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	5	20												10																
CO2	5	10	30						50																					
CO3		10	20						50																					
CO4													5	10	20															
CO5													5	10	15															
CO6														10	15															
Total (in %)	100						100						100						100											

## Syllabus

### DIGITAL SYSTEMS:

#### Number Systems and Codes:

Binary, Hexadecimal, Octal, BCD, 2421, Excess – 3, Gray

#### Digital Circuits and Gates:

Logical gates, TTL and MOS logic circuits

#### Combinational Logic:

Design and implementation of Half and Full adders, Encoders and decoders, Multiplexers and De-multiplexers

#### Sequential Logic:

R-S, J-K, and D Type Flip-Flops, Moore/Mealy models

#### Binary counters:

Ripple and synchronous types, Shift registers

### COMPUTER ORGANIZATION

#### Functional Units and Bus Structures

Basic operational concepts, Bus structures

#### Machine Instructions and Addressing Modes -Machine instructions, Addressing modes

#### Arithmetic Operations- Number representations, operations of signed numbers, Addition, Multiplication and division operation

#### Processing Unit and ALU -Concepts of processing unit, Execution of complete instruction, Multi bus organization, Arithmetic and Logic Unit (ALU)- Adders, Fast adders, Multiplication and Division circuits

#### Control Unit – Control units type, Hardwired control, Microprogrammed - Micro instructions, Micro program sequencing, Micro instructions types

#### Memory: Memory units, memory types - efficiency

#### Input and Output Organization -Accessing I/O devices, Interrupts, Direct Memory Access (DMA), Interface circuits

#### Advanced Processor Architecture- RISC, Pipelining, Super scalar processors, VLIW, Parallel and vector processors

## Reference Books

1. Leach D.P., "Digital Principles & Application", Tata McGraw Hill, 2014.
2. William Stallings, "Computer Organization and Architecture", Eleventh Edition, Pearson Education, 2022.
3. Carl Hamacher, Zvonko Vranesic, safwat Zaky, "Computer Organization and Embedded System", Sixth Edition, Tata McGraw Hill, 2022.
4. David A. Patterson, John L. Hennessy, "Computer Organization and Design", Fourth Edition, Morgan Kauffmann Publishers, 2020.

**Course Contents and Lecture Schedule**

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>1</b>	<b>Digital Systems - Introduction</b>	
1.1	Fundamental aspects of semiconductors	1
<b>2</b>	<b>Number Systems and Codes</b>	
2.1	Binary	1
2.2	Hexadecimal	1
2.3	BCD, Gray, Excess-3, 2421	1
<b>3</b>	<b>Digital Circuits ,Gates and Combinational Logic:</b>	
3.1	AND, OR, NOT gates	2
3.2	TTL and MOS logic circuits	2
3.3	Design and implementation of Half and Full adders	1
3.4	Encoders and decoders	1
3.5	Multiplexers and De-multiplexers	1
<b>4</b>	<b>Sequential Logic</b>	
4.1	R-S, J-K, and D type Flip-Flops	1
4.2	Binary counters: Ripple and synchronous types	1
4.3	Shift registers	1
<b>5</b>	<b>Computer architecture</b>	
5.1	Basic operational concepts and functional units	1
5.2	Bus structures	1
5.3	Machine instructions	1
5.4	Addressing modes	1
<b>6</b>	<b>Design of ALU</b>	
6.1	Number representations	1
6.2	Addition and subtraction of signed numbers	1
6.3	Design of fast adders	1
6.4	Multiplication of signed numbers	1
6.5	Fast multiplication	1
6.6	Integer division	1

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
6.7	Concepts of processing unit	1
6.8	Execution of complete instruction	1
6.9	Multi bus organization	1
<b>7</b>	<b>Control Unit and Memory</b>	
7.1	Hardwired control	1
7.2	Micro programmed control	1
7.3	Micro instructions	1
7.4	Micro program sequencing	1
7.5	Micro instructions with next address field and pre-fetching	1
7.6	RAM and ROM	2
7.7	Cache memories and Virtual Memory	2
<b>8</b>	<b>Input &amp; Output Organization</b>	
8.1	Accessing I/O devices	1
8.2	Interrupts	1
8.3	Direct Memory Access (DMA)	1
8.4	Interface circuits	1
<b>9</b>	<b>Advanced Processor Architecture</b>	
9.1	RISC	1
9.2	Pipelining	1
9.3	Super scalar processors, VLIW, Parallel and vector processors	2
	<b>Total</b>	<b>44</b>

**Course Designer(s):**

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23DS140

**PROBLEM SOLVING USING C PROGRAMMING**

Category L T P Credit  
FC 4 0 0 4

**Preamble**

The course aims to provide exposure to problem-solving through C programming. It aims to train the students on the basic concepts of the C-programming language with appropriate problems that reflects users' simple computing requirements

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Explain the basic constructs of programming languages	Understand
CO2	Given a computational problem, abstract the programming task involved with algorithm and flowchart	Apply
CO3	Write C programs with suitable operators and control flow statements for given user requirements.	Apply
CO4	Develop C programs with arrays and pointers in the context of functions with an understanding on the basic constructs of a programming language involving character and string functions	Apply
CO5	Apply the concept of structures and unions in developing application-oriented program in C	Apply
CO6	Develop C programs for user requirements with file operations	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	L	-	S	-	-	-	-	-	-	-	L	L
CO2	S	M	L	-	S	-	-	-	M	-	-	-	M	M
CO3	S	M	L	-	S	L	L	-	M	-	-	-	M	M
CO4	S	M	L	-	S	L	L	-	M	-	-	-	M	M
CO5	S	M	L	-	S	L	L	-	M	-	-	-	M	M
CO6	S	M	L	-	S	L	L	-	M	-	-	-	M	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	10	5																							5	5				
CO2	5	5	20						50																5		10			
CO3	5	10	20						25																	5	15			
CO4			20						25				5		20						40				5		15			
CO5													10	10	20						30					10	10			
CO6													5	10	20						30				5		10			
Total (in %)	100						100						100						100											

## Syllabus

**Computing and problem solving:** Fundamentals of Computing, Algorithms, Pseudocode and flowcharts. Role of programming languages, Memory, Variables, Values, Instructions, Programs, The Role of Programming Languages, Structured Programming.

**Types, operators and expressions:** Variables, Data types – constants – Declarations – Storage classes - Arithmetic, relational and logical operators - Type conversions – Increment, Decrement and bitwise operators – Assignment operators and expressions - conditional expressions – Precedence and order of evaluation – Practice problems

**Control flow:** Statements and blocks – Decision control structures – Looping control structures - Case control structures – Break and continue – Practice problems

**Functions and Program Structure:** Basics – Function Prototyping - Function Call by value - Function Call by reference – Recursion – Practice problems

**Pointers and Arrays:** Basic of pointers and arrays - Pointer Arithmetic - Pointers to array - Pointers to function - Multidimensional arrays and pointers –Practice problems

**Structures and Unions:** Motivation and examples – Structures and functions – Array of structures – Pointer to structures – Self-referential structures – Dynamic allocation - Table lookup – Unions – Bit fields - Practice problems

**I/O Management:** Input and output from console – Standard input – Formatted output – Variable length argument lists – Formatted input - File Management in C – File access – Error handling - Practice problems

## Reference Books

1. Byron S Gottfried, "Programming with C", 4th Edition, 2018
2. Yashavant Kanetkar, "Let Us C: Authentic guide to C programming language" , 18th Edition, BPB Publications, 2021
3. Sumitabha Das, "Computer Fundamentals and C Programming", McGraw Hill Education, 2018
4. Reema Thareja "Computer fundamentals and programming in C", Oxford University, Second edition, 2017
5. Brian W Kernighan & Dennis Ritchie, "The C programming language", 2nd Edition, Prentice Hall , 2015

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>Introduction to computing and problem solving</b>	
1.1	Fundamentals of Computing, Algorithms, Pseudo-code and flowcharts	1
1.2	Role of programming languages, Memory, Variables, Values, Instructions, Programs	1
1.3	Structured Programming	1
<b>2</b>	<b>Types, operators and expressions:</b>	
2.1	Variables and Data types , constants	1

Module No.	Topic	No. of Periods
2.2	Arithmetic, relational and logical operators	1
2.3	Type conversions	1
2.4	Increment, Decrement and bitwise operators	1
2.5	Assignment operators	1
2.6	conditional expressions	1
2.7	Precedence and order of evaluation	1
2.8	Practice problems	1
<b>3</b>	<b>Control flow:</b>	
3.1	Statements and blocks	1
3.2	Decision control structures	1
3.3	Looping control structures	1
3.4	Case control structures	1
3.5	Practice problems	1
<b>4</b>	<b>Functions and Program Structure:</b>	
4.1	Basics	1
4.2	Function Prototyping	1
4.3	Function Call by value	1
4.4	Function Call by reference	1
4.5	Recursion	1
4.6	Practice problems	1
<b>5</b>	<b>Pointers and Arrays:</b>	
5.1	Basic of pointers and arrays	1
5.2	Pointers to array	1
5.3	Pointers to function	1
5.4	Multidimensional arrays and pointers	1
5.5	Practice problems	1
<b>6</b>	<b>Structures and Unions:</b>	
6.1	Motivation and examples	1
6.2	Structures and functions	1
6.3	Array of structures	1
6.4	Self-referential structures	1
6.5	Pointer to structures	1
6.6	Dynamic allocation	1
6.7	Unions	1
6.8	Practice problems	2
<b>7</b>	<b>I/O Management:</b>	
7.1	Input and output from console	1

Module No.	Topic	No. of Periods
7.2	Standard input and Formatted output	1
7.3	Variable length argument lists	1
7.4	Formatted input	1
7.5	File Management in C	1
7.6	File access	1
7.7	Error handling	1
7.8	Practice problems	1
	Total	44

**Course Designer(s):**

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**23DS150**

**DISCRETE STRUCTURES**

Category L T P Credit  
FC 4 0 0 4

**Preamble**

This course will provide an outline to equivalent logical proposition for a real-world statement by applying predicates and quantifiers and to develop fundamental knowledge of combinatorics. This course also helps to apply the Laws using Lattices and Boolean algebra and prove the properties and construct the Automata to reduce the complexity of a model.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Prove implication problems using truth table method, truth table technique, rules of inference to obtain PCNF and PDNF of given logical expression.	Apply
CO2	Construct verbal arguments with predicates in symbolic form after validating them using inference	Apply
CO3	Represent the different types of relation in matrix, digraph and vice versa.	Apply
CO4	Prove using mathematical induction and problems using permutations and combinations.	Apply
CO5	Prove the properties of Lattices and Boolean algebra.	Apply
CO6	Construct DFA and NFA with given language considering the grammar	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1.	L	L	L	M	M	M	-	-	-	-	-	-	L	M
CO2.	S	S	M	M	-	M	-	-	-	-	-	-	L	S
CO3	S	S	M	M	-	-	-	-	-	-	-	-	L	S
CO4	S	S	M	S	M	M	-	-	-	-	-	-	L	S
CO5	S	S	M	S	M	-	-	-	-	-	-	-	L	S
CO6	S	S	L	L	M	M	-	-	-	-	-	-	L	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	5	10	20						40																	5	16			
CO2	5		20						30																2		8			
CO3		10	30						30																2	5	8			
CO4													5	10	30						40				2	5	12			
CO5													5	10	20						30				2	5	16			
CO6															20						30				2		10			
Total (in %)	100						100						100						100											

**Syllabus**

### **Mathematical logic:**

**Propositional Logic:** Introduction, Connectives. Tautology and Contradiction. Logical Equivalences and Logical Implications. Normal forms: CNF, DNF, PCNF, PDNF. Theory of Inferences - Rules of Inference – Consistency of premises. Validity by truth table technique.

**Predicate Calculus:** Predicates - Statement Function, Variables and Quantifiers, Theory of inferences of Predicate Calculus.

### **Set Theory:**

**Relations:** Relations, Binary relations, Equivalence relations and Partitions, Matrix representation of a relation, relation representation by graphs. Partially ordered set (PO Set), Hasse Diagram, LUB, GLB. **Functions:** Types, Composition of functions, Inverse of a function, Recursive functions.

### **Combinatorics:**

Mathematical Induction. Basics of Counting. Pigeonhole Principle. Permutations and Combinations. Recurrence relations. Principle of Inclusion and Exclusion.

### **Lattices & Boolean Algebra:**

Lattices – Principle of Duality - Properties of lattices, Sub lattices, Special lattices. Properties of Boolean algebra – Dual and Principle of Duality – Sub algebra. Boolean Homomorphism, Expressions and Functions.

### **Grammar and Automata Theory:**

**Grammar:** Phrase-Structure Grammars - Types of Grammar.

**Automata Theory:** Overview of Finite - State Machine, Input and Output strings of FSM, Finite State Automata, Pushdown Automaton, Turing Machine. Pumping Lemma.

### **Reference Books and Web Resources**

1. Kenneth H. Rosen, "Discrete mathematics and its applications", McGraw-Hill International Editions, 2021.
2. Trembly and Manohar, "Discrete mathematical structures with applications to Computer Science", Tata McGraw-Hill, 2017.
3. John E. Hopcraft, Rajeev Motwani, Jeffery D. Ullman, "Introduction to Automata Theory, Languages and Computation", Pearson Education, Asia, 3<sup>rd</sup> edition, 2013.
4. Judith L. Gersting, "Mathematical Structures for Computer Science", W.H. Freeman and Company, 2014.
5. <http://www.discrete-math-hub.com/resources-and-help.html#otherstuff>
6. [https://en.wikibooks.org/wiki/Discrete\\_Mathematics](https://en.wikibooks.org/wiki/Discrete_Mathematics)
7. <https://mathworld.wolfram.com/>

### **Course Contents and Lecture Schedule**

Module No.	Topic	No. of Periods
1	Mathematical logic	
1.1	Propositional Logic Introduction, Connectives	1
1.2	Tautology and Contradiction	1
1.3	Logical Equivalences and Logical Implications	1
1.4	Normal forms: CNF, DNF, PCNF, PDNF	2
1.5	Theory of Inferences - Rules of Inference – Consistency of premises. Validity by truth table technique	2
1.6	Predicate Calculus Predicates - Statement Function	1
1.7	Variables and Quantifiers	1
1.8	Theory of inferences of Predicate Calculus	2
2	Set Theory	
2.1	Relations: Binary relations, Equivalence relations and Partitions	1
2.2	Matrix representation of a relation, relation representation by graphs.	2
2.3	Partially ordered set (PO Set), Hasse Diagram, LUB, GLB.	2
2.4	Functions: Types, Composition of functions,	1
2.5	Inverse of a function, Recursive functions.	2
3	Combinatorics	
3.1	Mathematical Induction	1
3.2	Basics of Counting. Pigeonhole Principle.	2
3.3	Permutations and Combinations	2
3.4	Recurrence relations.	2
3.5	Principle of Inclusion and Exclusion.	2
4	Lattices & Boolean Algebra	
4.1	Lattices – Principle of Duality	1
4.2	Properties of lattices, Sub lattices, Special lattices	2

4.3	Properties of Boolean algebra – Dual and Principle of Duality	2
4.4	Sub algebra	1
4.5	Boolean Homomorphism, Expressions and Functions.	2
5	Grammar and Automata Theory	
5.1	Grammar: Phrase-Structure Grammars - Types of Grammar.	1
5.2	Automata Theory: Overview of Finite-state Machine	1
5.3	Input and Output strings of FSM	2
5.4	Finite State Automata, Pushdown Automaton	2
5.5	Turing Machine. Pumping Lemma.	2
	Total	44

**Course Designer(s):**

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 Dept. of Applied Mathematics and Computational Science.

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<b>23DS170</b>	<b>C PROGRAMMING LAB</b>	Category	L	T	P	Credit
		FC	0	0	4	2

**Preamble**

The course aims to provide exposure to problem-solving through C programming and involves a lab component which is designed to give the student hands-on experience with the concepts.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Implement C programs by choosing suitable datatypes and apply the concept of Control Structures to solve any given problem	Apply
CO2	Apply the concept of single and multi-dimensional arrays to solve problems related to searching, sorting and matrix operations.	Apply
CO3	Develop C programs using the concept of user-defined and recursive functions.	Apply
CO4	Develop C Programs by applying the concept of structures and unions	Apply
CO5	Implement C programs for demonstrating pointer concepts.	Apply
CO6	Implement C programs to create and process data files.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1.	S	S	S	L	L	L	L	-	L	L	L	L	L	L
CO2.	S	S	S	L	L	L	L	-	L	L	L	L	L	L
CO3	S	S	S	L	L	L	L	-	L	L	L	L	L	L
CO4	S	S	S	L	L	L	L	-	L	L	L	L	L	L
CO5.	S	S	S	L	L	L	L	-	L	L	L	L	L	L
CO6	S	M	M	L	L	L	L	-	L	L	L	L	L	L

S- Strong; M-Medium; L-Low

**List of Experiments**

1. Simple programs to demonstrate data types and storage classes
2. Familiarizing conditional, control and looping statements.
3. Usage of single and double dimensional arrays
4. Implementation of functions and recursive functions (call by value and reference)
5. Defining and handling structures, array of structures and union.
6. Implementation of pointers- operation on pointers and pointer arithmetic
7. Implementation of dynamic storage allocation.
8. Creating and processing data files.

### Reference Books

1. Yashavant Kanetkar, "Let Us C : Authentic guide to C programming language" , 18th Edition, BPB Publications, 2021.
2. Brian W Kernighan & Dennis Ritchie, "The C programming language", 2nd Edition, Prentice Hall ,2015 Chapters 1,2,3,4,5,6,7
3. H. M. Deitel, P. J. Deitel, "C: How to program", 7th edition, Pearson Education, 2010. Chapter 13
4. R. G. Dromey, "How to Solve It By Computer", 12 th Edition, Pearson Education, 2007
5. Darnell and Margolis, "ANSI C- A Systematic programming Approach" Narosa publications, 2010.

### Course Designer:

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**23DS180 PROFESSIONAL ENGLISH**

Category L T P Credit

EEC 0 2 2 3

**Preamble**

The course will equip the students to apply knowledge and language skills to thrive in the digital era where data and information are recognized as pivotal to the academic, professional, and social life.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Associate the meaning with its vocabulary	Understand
CO2	Adapt practical techniques to listen and comprehend various speeches with the required emphasis	Apply
CO3	Read for Comprehending various texts	Understand
CO4	Express their ideas clearly and effectively in both spoken and written communication	Apply
CO5	Explore critically and analytically about common social issues	Apply
CO6	Discover a spirit of inquiry	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1									S	S			L	L
CO2									S	S			L	L
CO3									S	S			L	L
CO4									S	S			L	L
CO5									S	S			L	L
CO6									S	S			L	L

S- Strong; M-Medium; L-Low

**Assessment Pattern****Internal**

**No Common Continuous Assessment Test (CAT) will be conducted.**

Students' performance will be continuously assessed in various classroom activities on Listening, Speaking, Reading and Writing for 60 marks as detailed below

Listening Test	-	10
Speaking Test- Presentation /Brainstorming	-	25
Written Test (Basic Grammar and Paragraph Writing)	-	25

**External (Practical)**

Listening	- 20
Speaking (Presentation)	- 20
Brain storming on Social Issues	- 10
Writing	- 50

(Vocabulary Building, Synonyms and Antonyms, One-word Substitutions, Idioms & Phrases, SV Agreement, Tenses, Voices, Prepositions, Adjectives and Adverbs)

**Syllabus**

**Module 1:** Basic Grammar – Subject Verb Agreement, Tenses / Voices, Preposition, Adjectives and Adverbs

**Module 2 :** Vocabulary Building- Synonyms and Antonyms, One Word Substitution, Idioms & Phrases

**Module 3 :** Listening Skill – Listening to various TED talks and Documentaries

**Module 4 :** Speaking Skill – SWOT Analysis, Elevator Pitch, Brainstorming on various social issues, Presentation Strategy, Presentation using various digital tools, Facing and handling difficult questions

**Module 5 :** Reading Skill – Skipping, skimming, scanning , Intensive reading, Reading comprehension  
Extensive Reading – ‘The Power of Positive Thinking’ by Norman Vincent Peale

**Module 6 :** Writing Skill - Coherent markers, Logical Sequence of Sentences and Paragraphs, Interpretation of Graphics, Writing Agenda & Minutes

**Reference Books and web resources**

1. Courseware on “Technical Communication for Scientists and Engineers”, IIT Bombay,2015.
2. Lewis, Norman. “How to Read better & Faster”. New Delhi: Binny Publishing House. 1978
3. McCarthy, Michael and Felicity O’Dell. “English vocabulary in use : 100 Units of Vocabulary reference and practice” in Cambridge: CUP. 1996
4. N.M. White, “ Unlock – Listening and Speaking Skills 1” Cambridge University Press, 2014
5. <https://www.ted.com/talks>
6. ”Elephant Whisperers” – Documentary, 2022
7. Online YouTube videos on Current Issues

**Extensive Reading** – Norman Vincent Peale, “The Power of Positive Thinking”, Readers’ Paradise, 2020

**Course Contents and Lecture Schedule**

Module No.	Topic	No. of Periods	
		Tutorials	Lab
1	<b>Basic Grammar</b>		1
1.1	Tenses / Voices	1	
1.2	Subject Verb Agreement	1	
1.3	Preposition	1	
1.4	Adjectives and Adverbs	1	

2	<b>Vocabulary Building</b>		2
2.1	Synonyms and Antonyms	1	
2.2	One word substitution	1	
2.3	Idioms and Phrases	2	
3	<b>Listening Skills</b>		
3.1	Listening to TED Talks and Documentaries		4
4	<b>Speaking Skills</b>		
4.1	SWOT Analysis	1	1
4.2	Elevator Pitch (Video Recording)	1	2
4.3	Brainstorming on various social issues	2	2
4.4	Presentation Strategy, Facing and handling difficult questions	2	2
4.5	Preparing presentation using various digital tools		4
5	<b>Reading Skills</b>		
5.1	Skimming, skipping and scanning , and Intensive Reading Skills	2	
5.2	Reading Comprehension	1	2
6	<b>Writing Skills</b>		
6.1	Coherent Markers, Logical Sequence of Sentences and Paragraphs	2	
6.2	Paragraph Writing	1	
6.3	Interpretation of Graphics	2	
6.4	Writing Agenda and Minutes	2	
	Practice Test on LSRW Skills		4
		24	24
	<b>Total</b>		<b>48 hours</b>

**Course Designer(s):**

1. Dr.A.Tamilselvi,  
Professor,  
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2. Dr.RS.Swarnalakshmi,  
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**23DS210**

**THEORY OF PROBABILITY**

Category L T P Credit  
FC 4 0 0 4

**Preamble**

This course will enable students to apply conditional probability, Bayes' theorem to solve real time problems and to learn the concepts of moment generating functions to discrete and continuous distributions and find the probability values for the defined distributions. Also, the course enables them to learn the concepts of Limit theorems and realize the applications of central limit theorem

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Apply the concepts of basics of probability, conditional Probability and Bayes' Theorem to solve the real world problems	Apply
CO2	Infer discrete and continuous random variables, probability mass function and density function, expectation and variance	Apply
CO3	Apply the concepts of discrete and continuous distributions as binomial, Bernoulli, Poisson, geometric, uniform, normal, exponential, Weibull, and gamma distributions to the real world problems	Apply
CO4	Apply joint probability distributions to find marginal and conditional distributions and also transformation of two random variables	Apply
CO5	Apply the concepts of random variables and distributions for the central limit theorem.	Apply
CO6	Apply the concept of reliability for system of independent Components and expected system lifetime.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1.	L	L	L	M	L	-	-	-	-	-	-	-	L	M
CO2.	S	S	M	M	L	-	-	-	-	-	-	-	L	S
CO3	S	S	M	M	L	-	-	-	-	-	-	-	L	S
CO4	S	S	M	S	-	-	-	-	-	-	-	-	L	S
CO5	S	S	M	S	-	-	-	-	-	-	-	-	L	S
CO6	S	S	L	L	-	-	-	-	-	-	-	-	L	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	4	10	20																											
CO2	4	10	20																											
CO3	2	10	20																											
CO4													4	10	30										35					
CO5													2	10	20										35					
CO6													4		20										30					
Total (in %)	100						100						100						100											

**Syllabus****Probability Basic Concepts:**

Introduction - Sample space and events – Basic Theorems - Conditional Probability – Baye’s Theorem - Independence.

**Random Variables:**

Discrete and continuous random variables - probability mass function and density function - Distribution function - Expectation and variance – Moments and Moment Generating functions - Sums of independent random variables.

**Theoretical Distributions:**

**Discrete:** Binomial, Bernoulli, Poisson and Geometric.

**Continuous:** Uniform, Normal, Exponential, Weibull, and Gamma.

**Two-Dimensional Random Variable:**

Joint probability distributions - Marginal and conditional distributions - Statistical independence - Conditional expectation — Transformation of two random variables.

**Central Limit Theorems:**

Limit theorems: Markov and Chebyshev inequalities, Law of Large numbers, Central Limit Theorem.

**Reliability:**

Introduction - Structure Functions - Reliability of Systems of Independent Components - System Life as a Function of Component Lives - Expected System Lifetime.

**Reference Books and Web Resources**

1. Sheldon M. Ross, “Introduction to Probability Models”, Academic Press, 12<sup>th</sup> edition 2019.
2. Saeed Ghahramani, “Fundamentals of Probability with Stochastic Processes”, Pearson Education, 4<sup>th</sup> edition 2019.
3. Jay L Devore, “Probability and Statistics for Engineering and Sciences”, Cengage Learning, 2015.
4. K. S. Trivedi, “Probability and Statistics with Queuing, Reliability and Computer Science Applications”, John Wiley & Sons, 2016.
5. <https://www.khanacademy.org/math/statistics-probability>
6. <https://www.probabilitycourse.com/>
7. <https://mathworld.wolfram.com/topics/ProbabilityandStatistics.html>

**Course Contents and Lecture Schedule**

Module No.	Topic	No. of Periods
1	Probability Basic Concepts:	
1.1	Introduction and Sample space and events	1
1.2	Axiomatic approach to probability – Basic theorems.	1
1.3	Conditional Probability	1
1.4	Law of multiplication	1
1.5	Law of total probability and Bayes Theorem	1
1.6	Independence	2
2	Random Variables:	

Module No.	Topic	No. of Periods
2.1	Discrete random variables	1
2.2	Continuous random variables	1
2.3	probability mass function and density function	2
2.4	distribution function	1
2.5	Expectation and variance	2
3	Theoretical Distributions:	
3.1	Discrete: Binomial, Bernoulli	2
3.2	Poisson, Geometric.	2
3.3	Continuous: Uniform, Normal	2
3.4	Exponential, Weibull, and Gamma.	2
4	Two Dimensional Random Variables:	
4.1	Joint probability distributions	1
4.2	Marginal and conditional distributions	2
4.3	Statistical independence	1
4.4	Conditional expectation	1
4.5	Transformation of two random variables.	2
5	Central Limit Theorems:	
5.1	Moments and moment generating functions	2
5.2	Sums of independent random variables	1
5.3	Limit theorems: Markov and Chebyshev inequalities	2
5.4	Law of Large numbers	1
5.5	Central Limit Theorem.	1
6	Reliability:	
6.1	Introduction - Structure Functions	2
6.2	Reliability of Systems of Independent Components	2
6.3	System Life as a Function of Component Lives	2
6.4	Expected System Lifetime	2
	Total	44

**Course Designer(s):**

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**23DS220**

**APPLIED STATISTICS**

Category L T P Credit

PCC 4 0 0 4

**Preamble**

The students will be able to learn the data analysis with the basics of graphs and tables, Test the hypothesis on mean, variance, proportion of small and large samples, for goodness of fit and independence of attributes and apply the concept of correlation, linear and nonlinear regressions to Data Science problems.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Learn the basic concepts of distributions, charts and various types of measures.	Understand
CO2	Apply the concepts of estimation and its type in mean, proportion and variance.	Apply
CO3	Demonstrate the concept of testing of hypothesis for small and large samples by using various tests like t-test, F-test, z-test and chi-square test	Apply
CO4	Apply the concept of Correlation and regressions to engineering problems Apply least square method in fitting linear and nonlinear regression curves.	Apply
CO5	Apply Multiple regression and correlation analysis, Inferences about population parameters and Modeling techniques.	Apply
CO6	Apply the concepts of Analysis of variance in real life problems	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	L	L	L	-	-	-	-	-	-	-	-	S	S
CO2	S	M	M	M	-	-	-	S	-	-	-	-	S	S
CO3	S	M	M	M	-	-	-	-	S	-	-	-	S	S
CO4	S	M	M	M	-	-	-	-	-	-	-	M	S	S
CO5	S	M	M	M	-	-	-	-	-	-	-	-	S	S
CO6	S	M	L	M	-	-	-	-	-	-	-	-	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	5	10																							2	5				
CO2	5	10	30						50																2	5	10			
CO3		10	30						50																2		20			
CO4													5	10	20					50				2	5	10				
CO5													5	10	20					25				2	5	10				
CO6														10	20					25					10	10				
Total (in %)	100						100						100						100											

## Syllabus

**DESCRIPTIVE STATISTICS:** Frequency distribution – Bar graphs and Pie charts – Histogram- Ogive – Simpson’s paradox – Measures of Location – Measures of Variability – Measures of distribution shape, relative location and detecting outliers – Exploratory Data analysis, Stem-and-leaf display – Measures of Association between two variables.

**STATISTICAL INFERENCE:** Sampling distribution - Estimation: Point estimation, interval estimation - Criteria of a good estimator – Interval estimation of mean, proportion, and variance (single sample and two samples).

**HYPOTHESIS TESTING:** General concepts - Errors in Hypothesis testing - One-and two-tailed tests - Tests concerning mean, proportion, and variance - Tests for Goodness of fit and independence of attributes.

**CORRELATION AND REGRESSION:** introduction - Estimation using the regression line - Correlation analysis -Limitations, errors, and caveats of using regression and correlation Analyzes - Multiple regression and correlation analysis - Inferences about population parameters – Modeling techniques.

**ANALYSIS OF VARIANCE:** Introduction to design of experiments, Analysis of variance.

**TOOLS:** d3.js, Excel, Tableau.

## Reference Books

1. Richard I. Levin. David S. Rubin, “Statistics for Management”, Pearson Education, 2017.
2. Ronald E. Walpole, Raymond H. Meyers and Sharon L. Meyers, “Probability and Statistics for Engineers and Scientists”, Pearson Education, 2019.
3. L. Devore, “Probability and Statistics for Engineering and Sciences”, Cengage Learning, 2018.
4. Anderson, Sweeney and Williams , “Statistics for business and economics”, Cengage Learning, 2014.
5. Saeed Ghahramani, “Fundamentals of Probability with Stochastic Processes”, Pearson Education, 2018.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>DESCRIPTIVE STATISTICS:</b>	
1.1	Frequency distribution	2
1.2	Bar graphs and Pie charts	1
1.3	Histogram- Ogive – Simpson’s paradox	2
1.4	Measures of Location – Measures of Variability	2
1.5	Measures of distribution shape, relative location and detecting outliers	2
1.6	Exploratory Data analysis, Stem-and-leaf display	2

1.7	Measures of Association between two variables	2
2	<b>STATISTICAL INFERENCE:</b>	
2.1	Sampling distribution - Estimation:	2
2.2	Point estimation, interval estimation	2
2.3	Criteria of a good estimator – Interval estimation of mean, proportion, and variance (single sample and two samples)	2
2.4	Maximum likelihood estimator	2
2.5	Hypothesis Testing: General concepts - Errors in Hypothesis testing	2
2.6	One-and two-tailed tests	2
2.7	Tests for Goodness of fit and independence of attributes	1
3	<b>CORRELATION AND REGRESSION:</b>	
3.1	Introduction - Estimation using the regression line	2
3.2	Correlation analysis -Limitations	2
3.3	Errors, and caveats of using regression and correlation	2
3.4	Multiple regression and correlation analysis	2
3.5	Inferences about population parameters	2
3.6	Modelling techniques	2
4	<b>ANALYSIS OF VARIANCE:</b>	1
4.1	Introduction to design of experiments,	2
4.2	Analysis of variance	1
4.3	Completely Randomized Design and Randomized Block Design.	1
5.0	Tools: d3.js, Excel, Tableau.	2
	<b>Total</b>	<b>45</b>

**Course Designers:**

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<b>23DS230</b>	<b>GRAPH THEORY</b>	Category	L	T	P	Credits
		FC	4	0	0	4

**Preamble**

This course will enable students to learn some basic types of graphs, connectivity and the complexity of the model and create the model to real time problem through graphs. It also enables to learn the random graphs and Ramsey numbers to apply in the social networks.

**Prerequisite**

- Nil

**Course Outcomes**

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

CO1	Explain the basic concepts of Graphs, walk, path, circuits and apply the concepts of Eulerian, Hamiltonian graphs in real life problems.	Apply
CO2	Explain the concepts of digraphs and relate the application of graph theory to computer science field.	Apply
CO3	Apply the concepts of trees and connectivity for network problems.	Apply
CO4	Apply the concepts of planarity and duality for solving geometric problems.	Apply
CO5	Use the matrix representation and solve minimum path finding problem.	Apply
CO6	Apply the concepts of Colouring, Covering and Domination for solving real time problems.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	S	S	-	-	L	-	-	-	-	-	L	M
CO2	S	S	S	S	-	-	-	-	-	-	-	-	L	S
CO3	M	M	S	S	-	-	-	-	-	-	-	-	L	S
CO4	S	S	S	S	-	-	M	-	-	-	-	-	L	S
CO5	S	S	S	S	-	-	M	-	-	-	-	-	L	S
CO6	M	S	S	S	-	-	M	-	-	-	-	-	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
<b>Bloom's level</b>	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	4	10	20						30																4	5	10			
CO2	2	10	20						40																2	-	20			
CO3	4	10	20						30																2	5	10			
CO4													4	10	20									2	-	10				
CO5													2	10	20									-	5	10				
CO6													4	10	20									-	5	10				
Total (in %)	100						100						100						100											

**Syllabus**

**INTRODUCTION:** Graphs and its Applications, Finite and Infinite Graphs, Incidence and Degree, Isolated Vertex, Pendent Vertex and Null Graph. Isomorphism, Sub graphs, Walks, Paths and Circuits, Connected and Disconnected Graphs and Components, Euler Graphs, Hamiltonian Paths and Circuits, Travelling Salesman Problem.

**DIRECTED GRAPHS:** Digraphs and Binary Relations Directed Paths and Connectedness, Euler Digraphs, Trees with Directed Edges, Acyclic Digraphs and Decyclization.

**TREES AND FUNDAMENTAL CIRCUITS:** Trees and its Properties, Distance and Centers, Rooted and Binary Trees, Spanning Trees. **Cut-Sets and Cut-Vertices:** Properties, Connectivity and Separability, Network Flows, 1-Isomorphism, 2-Isomorphism.

**PLANAR AND DUAL GRAPHS:** Planar Graphs, Kuratowski's Two Graphs, Combinatorial and Geometric Graphs and its Dual.

**MATRIX REPRESENTATION OF GRAPHS:** Incidence Matrix, Matrix - Path, Adjacency, Cut-Set, Application to a Switching Network.

**COLORING, COVERING AND PARTITIONING:** Chromatic Number, Partitioning, Polynomial, Matching, Coverings, Four Colour Problem.

**Reference Books and Web Resources:**

1. Narsingh Deo, "Graph Theory with Applications to Engineering and Computer Science", 2016.
2. Douglas B. West, "Graph Theory", Pearson Education, 2015.
3. Bondy J.A. and Murty U.S.R, "Graph Theory", Springer, 2013.
4. Balakrishnan R and Ranganathan K, "A Textbook of Graph Theory", Springer-Verlag, 2012.
5. <https://www.graphclasses.org/index.html>
6. <https://t5k.org/graph/>

**Course Contents and Lecture Schedule**

Module No.	Topic	No. of Periods
<b>1</b>	<b>Introduction</b>	
1.1	Basic Definitions of graphs with examples	3
1.2	Paths and Circuits concepts with example	3
1.3	Classify the Eulerian and Hamiltonian graphs	3
<b>2</b>	<b>DIRECTED GRAPHS</b>	
2.1	Digraphs and Binary Relations	2
2.2	Directed Paths and Connectedness	2
2.3	Trees with Directed Edges, Acyclic Digraphs and Decyclization.	3
<b>3</b>	<b>TREES &amp; CONNECTIVITY</b>	

Module No.	Topic	No. of Periods
3.1	Trees and its Properties	2
3.2	Concepts of Rooted, Binary Trees and Spanning Trees	2
3.3	Concepts of Connectivity and Separability	2
3.4	Explanation of Network Flows, 1-Isomorphism, 2-Isomorphism	2
<b>4</b>	<b>PLANARITY &amp; DUALITY</b>	
4.1	Concepts of Planar Graphs and its applications.	2
4.2	Kuratowski's algorithm of two Graphs	2
4.3	Classify the Geometric Graphs and it's Dual	2
<b>5</b>	<b>MATRIX REPRESENTATION</b>	
5.1	Incidence Matrix, Matrix - Path	2
5.2	Adjacency & Cut-set	2
5.3	Application to a Switching Network	2
<b>6</b>	<b>COLORING, COVERING AND PARTITIONING</b>	
6.1	Concepts of Coloring.	2
6.2	Chromatic Number, Four colour theorem.	2
6.3	Matchings & Coverings and its applications.	2
6.4	Partitioning & Polynomial of graphs.	2
	Total	44

**Course Designer(s):**

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**23DS240 OBJECT ORIENTED PROGRAMMING**

Category L T P Credit  
PCC 4 0 0 4

**Preamble**

To provide sound knowledge on basic and advanced concepts of Object-Oriented programming and apply them in developing industrial strength software applications.

**Prerequisite**

- 23DS140 Problem Solving Using C programming.

**Course Outcomes**

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

CO1	Explain the differences between structural and object oriented programming	Understand
CO2	Identify classes, relevant data representations, operations and interactions for the entities in a given problem description.	Apply
CO3	Develop object-oriented programs by applying Abstraction, Encapsulation and Information hiding.	Apply
CO4	Practice code reusability and extensibility by means of Inheritance and Polymorphism.	Apply
CO5	Illustrate the use of templates and template libraries for a given user requirement	Apply
CO6	Write C++ programs with I/O streams and file objects for a given user requirement	Apply
CO7	Select suitable object-oriented programming constructs for developing real time applications.	Analyze

**Mapping with Programme Outcomes**

Cos	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1.	S	M	S	-	L	-	-	-	-	-	-	-	M	L
CO2.	S	S	S	L	L	-	-	-	-	-	-	-	S	M
CO3	S	S	S	M	M	L	-	-	L	-	-	-	S	M
CO4	S	S	S	M	M	L	-	L	M	-	L	-	S	M
CO5	S	S	S	M	M	-	-	-	-	-	-	-	S	M
CO6	S	S	S	M	M	-	-	L	-	-	-	-	S	M
CO7	S	S	S	M	M	L	-	L	M	-	M	L	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	5	10																							4	5				
CO2	5	10	20																						4		10			
CO3		10	20						50																4	5	10			
CO4			20						50																4	5	10			
CO5													5	10	20					25				2		10				
CO6													5	10	20					25				2	5	10				
CO7														10		20						50						10		
Total (in %)	100						100						100						100						100					

## Syllabus

**Principles of Object Oriented Programming:** Software crisis Software Evolution - Procedure Oriented Programming - Object Oriented Programming Paradigm - Basic Concepts and Benefits of OOP - Object Oriented Programming Language - Application of OOP.

**Structure of C++:** Tokens, Expressions and Control Structures - Operators in C++ - Manipulators.

**Functions in C++:** Function Prototyping - Call by Reference - Return by reference - Inline functions - Default, Const Arguments - Function Overloading

**Classes and Objects:** Member functions - Nesting of Member functions - Private member functions - Memory allocation for Objects - Static data members - Static Member Functions - Arrays of Objects - Objects as Function Arguments - Friend Functions - Returning Objects - Const Member functions - Pointers to Members – Virtual functions.

**Constructors:** Parameterized Constructors - Multiple Constructors in a Class - Constructors with Default Arguments - Dynamic Initialization of Objects - Copy and Dynamic Constructors – Destructor overloading.

**Operator Overloading:** Overloading Unary and Binary Operators - Overloading Binary Operators using Friend functions – Operator Type conversion.

**Inheritance:** Defining Derived Classes - Single Inheritance - Making a Private Member Inheritable - Multiple Inheritance - Hierarchical Inheritance - Hybrid Inheritance - Virtual Base Classes - Abstract Classes - Constructors in Derived Classes - Member Classes - Nesting of Classes – Composition – Aggregation.

**Polymorphism:** Basics of polymorphism – Types of polymorphism - Compile and Run Time Polymorphism - Virtual function – Object Slicing – Virtual Destructor – Dynamic binding.

**Templates & Exception Handling:** Introduction to Templates, Generic Functions and Generic Classes – Exception Handling – STL Components.

**Streams:** String I/O - Character I/O - Object I/O - I/O with multiple Objects - File pointers - Disk I/O with member functions.

## Reference Books

1. Bjarne Stroustrup, "The C++ Programming Language", Pearson Education, 2013.
2. Yashavant P. Kanetkar, "Let Us C++", BPB Publications, 2020.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>Principles of Object Oriented Programming</b>	
1.1	Software crisis Software Evolution - Procedure Oriented Programming - Object Oriented Programming Paradigm	2
1.2	Basic Concepts and Benefits of OOP -Object Oriented Programming Language	2
1.3	Application of OOP	1
<b>2</b>	<b>Structure of C++</b>	

Module No.	Topic	No. of Periods
2.1	Tokens, Expressions and Control Structures	1
2.2	Operators in C++ - Manipulators	1
<b>3</b>	<b>Functions in C++</b>	
3.1	Function Prototyping - Call by Reference - Return by reference	1
3.2	Inline functions - Default, Const Arguments	1
3.3	Function Overloading.	1
<b>4</b>	<b>Classes and Objects</b>	
4.1	Member functions - Nesting of Member functions - Private member functions	2
4.2	Memory allocation for Objects - Static data members - Static Member Functions	2
4.3	Arrays of Objects - Objects as Function Arguments - Friend Functions	1
4.4	Returning Objects - Const Member functions - Pointers to Members- virtual functions.	1
<b>5</b>	<b>Constructors</b>	
5.1	Parameterized Constructors - Multiple Constructors in a Class	1
5.2	Constructors with Default Arguments - Dynamic Initialization of Objects	2
5.3	Copy and Dynamic Constructors – Destructors overloading.	2
<b>6</b>	<b>Operator Overloading</b>	
6.1	Overloading Unary and Binary Operators	1
6.2	Overloading Binary Operators using Friend functions	2
6.3	Operator Type conversion	1
<b>7</b>	<b>Inheritance</b>	
7.1	Defining Derived Classes - Single Inheritance - Making a Private Member Inheritable	2
7.2	Multiple Inheritance - Hierarchical Inheritance - Hybrid Inheritance	1
7.3	Virtual Base Classes - Abstract Classes - Constructors in Derived Classes	1
7.4	Member Classes - Nesting of Classes – Composition – Aggregation	2
<b>8</b>	<b>Polymorphism</b>	
8.1	Basics of polymorphism – Types of polymorphism	2

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
8.2	Compile and Run Time Polymorphism - Virtual function	1
8.3	Object Slicing – Virtual Destructor – Dynamic binding	2
<b>9</b>	<b>Templates &amp; Exception Handling</b>	
9.1	Introduction to Templates, Generic Functions and Generic Classes	2
9.2	Exception Handling – STL components.	2
<b>10</b>	<b>Streams</b>	
10.1	String I/O -Character I/O - Object I/O	1
10.2	I/O with multiple Objects - File pointers	2
10.3	Disk I/O with member functions.	1
	<b>Total</b>	<b>44</b>

**Course Designer(s)**

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**23DS250 ORGANIZATIONAL THEORY AND BEHAVIOUR** Category L T P Credit  
PCC 3 0 0 3

**Preamble**

To learn challenges and opportunities in organizations from a behavioral science perspective.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Develop an Organisational Behaviour model for any type of Organization.	Apply
CO2	Develop Managerial skills for Individual Behaviours.	Apply
CO3	Develop the quality of Leadership.	Apply
CO4	Analyze the Common biases and eradication in Decision Making Process.	Analyze
CO5	Adapt to the organizational culture.	Apply
CO6	Analyze how to manage the Stress during a job.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	-	-	-	-	-	-	-	-	-	-	L	-
CO2	S	M	-	-	-	-	-	-	S	-	-	-	L	-
CO3			-	-	-	-	-	-	-	-	M	-	L	-
CO4	S		M	-	-	-	-	M	-	-	S	-	-	L
CO5	S	S	M	-	-	-	-	-	-	-	-	-	-	L
CO6	S	S	L	-	-	-	-	-	-	-	-	-	L	L

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	5	10	20						50																2	5				
CO2	5	10	20						25																2	5	10			
CO3		10	20						25																2	5	10			
CO4													10	10	20							30		2	5		20			
CO5													5	10	15							40		2	5	10				
CO6													5	10		15						30		5		10				
Total (in %)	100						100						100						100						100					

## Syllabus

**Organizational Behavior:** Introduction to Organizational Behavior(OB), Contributing disciplines to OB Field, challenges and opportunities for OB, Developing an OB Model, Foundation of Individual Behavior, Ability - Learning. Values, Attitudes and Types of Attitudes.

**Job satisfaction-** Measuring Job satisfaction, Effect of Job satisfaction on employee performance.

**Personality and Values:** Personality determinants, Achieving personality fit, Factors Influencing perception, Attribution Theory, Perception / Individual Decision Making: Ethics in Decision Making. Motivation, Management by Objectives.

**Understanding work teams:** Teams Vs Groups – Types of Teams, Creating Effective Teams – Turning Individuals into Team Players.

**Communication:** Functions of Communication, Communication Process – Direction of communication, Interpersonal and Organizational communication, Barriers of effective communication, Current issues in Communication.

**Leadership:** Leadership – Meaning, Trait Theories – Behavioral Theories, Contingency Theories, Contemporary issues in Leadership.

**Organizational culture and Stress Management:** Meaning – Creating and sustaining culture, How employees learn culture, creating an ethical organizational culture, creating a customer responsive culture, spirituality and organizational culture, Work stress and its Management.

## Reference Books

1. Stephen P. Robbins, "Organisational Behaviour", 11/e, Pearson, 17<sup>th</sup> Edition 2022 (Chapters 1,3,5, 6, 10,11,12, 16,17)
2. Uma Sekaran, "Organisational Behaviour", 2/e, Tata McGraw-Hill Publishing Company Ltd., New Delhi, 2010.
3. Sharma, R.A, "Organisational Theory and Behaviour", 2/e, Tata McGraw-Hill Ltd., New Delhi, 2012.

## Course Contents and Lecture Schedule

Module .No	Topics	No. of Lectures
1	<b>Organizational Behavior</b>	
1.1	Introduction to Organizational Behavior (OB)	1
1.2	Contributing disciplines to OB Field	1
1.3	Challenges and opportunities for OB	1
1.4	Developing an OB Model	1
1.5	Foundation of Individual Behavior	1
1.6	Ability - Learning. Values, Attitudes and Types of Attitudes	1
	<b>Job satisfaction</b>	
1.7	Measuring Job satisfaction.	1

1.8	Effect of Job satisfaction on employee performance.	1
2	<b>Personality and Values</b>	
2.1	Personality – Personality determinants	1
2.2	Achieving personality fit	1
2.3	Factors Influencing perception	1
2.4	Attribution Theory	1
	<b>Perception / Individual Decision Making</b>	
2.5	Ethics in Decision Making. Motivation	1
2.6	Management by Objectives	1
3	<b>Understanding work teams</b>	
3.1	Teams Vs Groups – Types of Teams	1
3.2	Creating Effective Teams – Turning Individuals into Team Players	1
	Communication	
3.3	Functions of Communication	1
3.4	Communication Process – Direction of communication	1
3.5	Interpersonal and Organizational communication	1
4	<b>Leadership</b>	1
4.1	Leadership – Meaning	2
4.2	Trait Theories – Behavioral Theories	1
4.3	Contingency Theories	2
4.4	Contemporary issues in Leadership	2
5	<b>Organizational culture</b>	1
5.1	Organizational culture: Meaning – Creating and sustaining culture	1
5.2	How employees learn culture	1
5.3	creating an ethical organizational culture	1
5.4	creating a customer responsive culture	1
5.6	Approaches to Managing organizational change	1
5.7	Creating a Culture	1
5.8	Change in Business	1
5.9	Work stress and its Management	1
	<b>Total</b>	<b>36</b>

**Course Designers:**

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<b>23DS270</b>	<b>PYTHON PROGRAMMING AND APPLIED STATISTICS LAB</b>	Category	L	T	P	Credit
		PCC	0	0	6	3

**Preamble**

The purpose of this course is to introduce students to the field of programming using Python for statistically displaying and describing data, including the normal curve, regression, probability, statistical inference, confidence intervals, and hypothesis tests, with applications in the real world.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Make use of branching, looping, strings, and function concepts to develop Python programmes for the given problem.	Apply
CO2	Implement object-oriented programming concepts and file handling in Python.	Apply
CO3	Design a basic data structure using Python programming constructs and its packages.	Apply
CO4	Interpret the usage of packages and libraries in Python programming for exploratory data analysis and data cleaning.	Apply
CO5	Perform the regression-correlation and statistical inference analyses.	Apply
CO6	Perform hypothesis testing (z test, t test, F test, and chi squared test).	Apply

**Mapping with Programme Outcomes**

Cos	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1.	S	M	S	L	L	-	-	-	-	-	-	-	S	M
CO2.	S	S	S	L	L	-	-	-	-	-	L	-	S	M
CO3	S	S	S	L	L	-	-	-	L	-	L	-	S	M
CO4	S	S	S	M	M	L	-	L	M	-	L	L	S	S
CO5	S	S	S	M	M	L	-	L	M	-	-	-	S	S
CO6	S	S	S	M	M	L	-	L	M	-	M	L	S	S

S- Strong; M-Medium; L-Low

## List of Experiments

### Develop Python Programs for

1. Basic looping and branching programs.
2. Programs on Dictionary, tuples and Nested Lists.
3. User defined functions.
4. String programs.
5. Classes and Objects.
6. File handling with exceptions.
7. Basic Data structures design using python.

### Solve the Statistical Problems using Python:

1. Exploratory Data Analysis and Data Cleaning- Pre-defined libraries & packages.
2. Perform calculations that measure the central tendency – Mean, Median, Mode and dispersion of data – Percentile, Range, Standard Deviation.
3. Visualizing Dispersion & Frequency (Boxplot, Histogram, Bar chart, Pie chart, Line charts, etc.)
4. Implementation of measures of Skewness, moments and kurtosis.
5. Determination of point and interval estimates.
6. Calculate and interpret the correlation coefficient of the two variables. (pearson, Spearman)
7. Perform paired t test, F test and Chi square test.
8. Hypothesis testing & ANOVA.
9. Solve linear regression line using python for the problem.

## Reference Books

1. John V.Guttag, "Introduction to Computation and Programming Using Python : With Application to Understanding Data", Prentice-Hall International publishers, Second Edition, 2017.
2. Yashavant Kanetkar, Aditya Kanetkar, "Let us Python, BPB publication, 1st Edition, 2019.
3. I. Levin Richard, H. Siddiqui Masood, S. Rubin David, Rastogi Sanjay, "Statistics for Management", Pearson Education, 2017.
4. Ronald E. Walpole, Raymond H. Meyers and Sharon L. Meyers, "Probability and Statistics for Engineers and Scientists", Pearson Education, 2014.

## Course Designer(s)

Mrs. R. Saraswathi Meena,  
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**23DS280****OBJECT ORIENTED PROGRAMMING  
LAB**Category L T P Credit  
PCC 0 0 4 2**Preamble**

This course is intended for the students to learn Object Oriented Programming and the design of computer solutions with the programming principles. The course emphasizes on practicing OOP concepts, Functions, Polymorphism, Inheritance and I/O and enables them to apply Object Oriented Programming approach to programming.

**Prerequisite**

- 23DS170: C Programming LAB

**Course Outcomes**

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

CO1	Illustrate the basic concepts of object oriented programming for building object based applications	Apply
CO2	Develop programs using object-oriented concepts like abstraction, encapsulation, polymorphism and inheritance to solve the given problem.	Apply
CO3	Demonstrate overloading of methods and operators with appropriate object-oriented programming constructs for different user specifications with suitable programs	Apply
CO4	Implement templates with standard template libraries and create user defined templates.	Apply
CO5	Develop programs for creating packages and handle exceptions for any real-time requirements	Apply
CO6	Create object-oriented programs to demonstrate file handling	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1:	S	M	L	-	L	L	-	-	-	-	-	-	M	M
CO2:	S	M	L	-	L	L	-	-	-	-	-	-	M	M
CO3:	S	M	L	-	L	L	-	-	-	-	-	-	M	M
CO4:	S	M	L	-	L	L	-	-	-	-	-	-	M	M
CO5:	S	M	L	-	L	M	-	-	-	-	-	-	M	M
CO6:	S	M	L	-	L	M	-	-	-	-	-	-	M	M

S- Strong; M-Medium; L-Low

### List of Experiments

Develop programs in C++ for different user specification using the below constructs in the context of classes and objects:

1. Implementation of classes and objects
2. Static member data and methods
3. Objects as Function Arguments and return types
4. Constructor types and its overloading
5. Method Overloading and method overriding with base class and derived class
6. Operator Overloading: Unary and Binary using member function & non-member Function.
7. Aggregation, composition and inheritance
8. Friend Function and Friend Class
9. Use of pointers within classes (this pointer, call and return of objects)
10. Abstract class, Virtual function and creating packages
11. Exception handling
12. File handling
13. Templates and STL Components

Mini project

### Reference Books

1. H.M. Deitel and P.J. Deitel, C “How to program Introducing C++ and Java”, Fourth Edition, Pearson Prentice Hall, (2012), Reprint 2020.
2. Yashavant P. Kanetkar, “Let Us C++”, BPB Publications, 2020.

### Course Designer(s):

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<b>23DS310</b>	<b>PARTIAL DIFFERENTIAL EQUATIONS AND TRANSFORMS</b>	Category	L	T	P	Credit
		FC	4	0	0	4

**Preamble**

This course will enable students to mathematically formulate practical problems in terms of partial differential equations, solve them and physically interpret the results using transformations.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Apply the concept of PDE for certain practical problems to solve them and physically interpret the results.	Apply
CO2	Apply Z transform technique to find the inverse Z transforms to solve the given ordinary differential equation and modelling.	Apply
CO3	Apply Laplace transform technique to find the inverse Laplace, solve the given ordinary differential equation and integral equations.	Apply
CO4	Express the periodic functions arising in the study of engineering problems as Fourier series of Sines and Cosines.	Apply
CO5	Obtain the Fourier series expansion in engineering problems using half range sine and cosine series.	Apply
CO6	Solve some of the well-known integral transforms (like Fourier, Fourier Sine and Cosine) and properties. Also Apply the Fourier concepts in solving Discrete and Fast Fourier transforms.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	S	S	S	M	L	L	-	-	-	-	M	L	S
CO2	M	S	M	S	L	L	L	-	-	-	-	L	L	S
CO3	S	S	M	S	M	L	L	-	-	-	-	M	L	S
CO4	M	S	M	S	L	L	L	-	-	-	-	M	L	S
CO5	S	S	M	S	L	L	L	-	-	-	-	M	L	S
CO6	S	S	M	S	L	L	L	-	-	-	-	M	L	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	4	10	20						30																2	5	10			
CO2	2	10	20						40																2	5	10			
CO3	4	-	30						30																2	-	20			
CO4													4	10	20									30						
CO5													2	10	20									30						
CO6													4	-	30									40						
Total (in %)	100						100						100						100											

## Syllabus

**PARTIAL DIFFERENTIAL EQUATIONS:** Formation of partial differential equations by elimination of arbitrary constants and functions – Solution of standard types of first order PDE – Lagrange's linear equation – Linear PDE of second and higher order with constant coefficients.

**Z-TRANSFORM:** Z - transform of standard functions - Properties - Inverse Z-transform — Convolution Theorem – Introduction to Difference operators - Formation and Solution of difference equations.

**LAPLACE TRANSFORM:** Transforms of Standard Functions - Transforms of derivatives and integrals - Transform of Unit step function and Dirac's delta function – Transforms of Periodic functions - Inverse Laplace transform - Partial Fraction - Convolution Theorem - Method of solving ordinary linear differential equations with constant coefficient and solving integral equations.

**FOURIER SERIES:** Dirichlet's conditions, statement of Fourier theorem, Fourier coefficients, Even and odd functions, change of scale, Half-range sine and cosine series, RMS value, Parseval's theorem.

**FOURIER TRANSFORM:** Fourier integrals - Fourier transform - Fourier sine and cosine transform - Transforms of standard functions - Properties, Convolution theorem – Discrete Fourier and Fast Fourier Transforms – Computation of inverse DFT - Properties of DFT.

## Reference Books and web resources

1. Erwin Kreyszig, "Advanced Engineering Mathematics", John Wiley & Sons, 10<sup>th</sup> edition 2014.
2. Ray Wylie C, Louis C Barret," Advanced Engineering Mathematics", McGraw Hill, 2013.
3. Roland E. Thomas and Albert J. Rosa, "The Design and Analysis of Linear Circuits", John Wiley & Sons, 2016.
4. Michael D. Greenberg, "Advanced Engineering Mathematics", Pearson Education, 2014.
5. <https://www.khanacademy.org/math/differential-equations/laplace-transform>
6. <https://tutorial.math.lamar.edu/Classes/DE/DE.aspx>

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>PARTIAL DIFFERENTIAL EQUATIONS</b>	
1.1	Formation of partial differential equations by elimination of arbitrary constants and functions.	2
1.2	Solution of standard types of first order PDE.	2
1.3	Lagrange's linear equation – Linear PDE of second and higher order with constant coefficients.	2
<b>2</b>	<b>Z-Transform</b>	
2.1	Z - transform of standard functions	2
2.2	Properties of Z-transform	2
2.3	Inverse Z-transform & Convolution Theorem	2
2.4	Introduction to difference operator	1
2.5	Formation and Solution of difference equations.	2
<b>3</b>	<b>Laplace Transform</b>	
3.1	Transforms of Standard Functions, Transforms of derivatives and integrals	2

Module No.	Topic	No. of Periods
3.2	Transform of Unit step function and Dirac's delta function - Transforms of Periodic functions	2
3.3	Inverse Laplace transform, Partial Fraction, Convolution Theorem	2
3.4	Method of solving ordinary linear differential equations with constant coefficient and solving integral equations.	2
<b>4</b>	<b>Fourier Series</b>	
4.1	Dirichlet's conditions, statement of Fourier theorem	3
4.2	Fourier coefficients, Even and odd functions	2
4.3	change of scale	2
4.4	Half-range sine and cosine series	2
4.5	RMS value, Parseval's theorem	2
<b>5</b>	<b>Fourier Transform</b>	
5.1	Fourier integrals - Fourier transform	2
5.2	Fourier sine and cosine transform	2
5.3	Transforms of standard functions - Properties, Convolution theorem	2
5.4	Discrete Fourier and Fast Fourier Transforms	2
5.5	Computation of inverse DFT, properties of DFT	2
	<b>Total</b>	<b>44</b>

**Course Designer(s):**

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**23DS320****ABSTRACT ALGEBRA**

Category	L	T	P	Credit
FC	4	0	0	4

**Preamble**

This course will enable students to learn the basic concepts of algebraic structures and groups, apply the concepts of normal subgroups and Cayley's theorem, model the concepts of Coding of Binary information and Error detection and learn the concepts of rings, fields and unique factorization theorem

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Apply the concepts of Groups, subgroups and its properties with Permutation.	Apply
CO2	Apply the concept of groups to learn about normal subgroups and fundamental theorem of group homomorphism	Apply
CO3	Apply the concept of groups to learn about coding of binary information, decoding, error correction and residue arithmetic	Apply
CO4	Apply the concepts of Rings to Learn its properties and different types of rings.	Apply
CO5	Apply the concept of rings to find factorization of polynomials and primitive polynomials	Apply
CO6	Apply the concept of fields to find the structure of finite field and GF	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	L	L	L	M	-	-	-	-	-	-	-	-	L	M
CO2	S	S	M	M	-	-	-	-	-	-	-	-	L	S
CO3	S	S	M	M	-	-	-	-	-	-	-	-	L	S
CO4	S	S	M	S	-	-	-	-	-	-	-	-	L	S
CO5	S	S	M	S	-	-	-	-	-	-	-	-	L	S
CO6	S	S	L	L	-	-	-	-	-	-	-	-	L	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	4	10	20						30																2	5	8			
CO2	4	10	30						35																2	5	16			
CO3	2	10	10						35																2		8			
CO4													4	10	30							35			2	5	14			
CO5													2	10	20							35			2	5	14			
CO6													4		20							30					10			
Total (in %)	100						100						100						100											

## Syllabus

### GROUPS:

Groups - Definition and Example, Properties of Groups, Permutation Groups, Symmetric Groups, Subgroups, Cyclic Groups, Cosets and Lagrange's theorem, Normal subgroups, Isomorphism, Homomorphism, Automorphism, Cayley's theorem, Factor group, Fundamental theorem of group homomorphism.

### CODING THEORY AND RESIDUE ARITHMETIC:

Coding of Binary information and Error detection – Group codes – Decoding and Error correction, Overview of Residue arithmetic.

### RINGS:

Definition and Properties – Subrings, Ring of Quaternions, Homomorphism, Ideals and Quotient Rings, Integral domain, Euclidean ring - Unique factorization theorem, Polynomials Rings – Properties, Division Algorithm, Factorization of Polynomials – Primitive polynomials.

### FIELDS:

Definition – Subfields - Finite fields – Galois Field.

## Reference Books and web resources

1. Joseph A. Gallian, "Contemporary Abstract Algebra", Brooks/Cole, 9<sup>th</sup> edition 2018.
2. Herstein I. N., "Topics in Algebra", John Wiley & Sons, 2012.
3. Tremblay J. P. and Manohar R., "Discrete Mathematical Structures with Applications to Computer Science", Tata McGrawHill, 2017.
4. Ralph P. Grimaldi and Ramana B. V., "Discrete and Combinatorial Mathematics: An Applied Introduction", Pearson Education, 5<sup>th</sup> edition 2019.
5. <https://mathworld.wolfram.com/AbstractAlgebra.html>

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	Groups	
1.1	Groups - Definition and Example	1
1.2	Properties of Groups	1
1.3	Permutation Groups, Symmetric Groups	2
1.4	Subgroups	2
1.5	Cyclic Groups	2
1.6	Cosets and Lagrange's theorem	2
1.7	Normal subgroups	1
1.8	Isomorphism	2
1.9	Homomorphism, Automorphism, Cayley's theorem	2
1.10	Factor group – Fundamental theorem of group	2

Module No.	Topic	No. of Periods
	homomorphism	
2	CODING THEORY AND RESIDUE ARITHMETIC:	
2.1	Coding of Binary information and Error detection	2
2.2	Group codes	2
2.3	Decoding and Error correction	2
2.4	Overview of residue arithmetic	3
3	RINGS:	
3.1	Definition and Properties	1
3.2	Subrings, Ring of Quaternions	2
3.3	Homomorphism, Ideals and Quotient Rings	2
3.4	Integral domain, Euclidean ring	2
3.5	Unique factorization theorem, Polynomials Rings – Properties	2
3.6	Division Algorithm, Factorization of Polynomials – Primitive polynomials	2
4	FIELDS:	
4.1	Definition	1
4.2	Subfields	2
4.3	Finite fields	2
4.4	Galois Field	2
	<b>Total</b>	<b>44</b>

**Course Designer(s):**

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**23DS330****DATA STRUCTURES**

Category L T P Credit

PCC 4 0 0 4

**Preamble**

This course aims at providing a deep understanding of various data structures, their operations and enables the students to identify appropriate data structures and design suitable algorithms for real world user requirements.

**Prerequisite**

- 23DS140 Problem Solving Using C programming

**Course Outcomes**

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

CO1	Explain the organization and operations of different data structures with an understanding of their efficiency	Understand
CO2	Demonstrate the operations of linear data structures (Array, Linked list, stack and queues)	Apply
CO3	Demonstrate the operations of non-linear data structures (Trees, Graph)	Apply
CO4	Demonstrate search and sort algorithms with an understanding of their analysis	Apply
CO5	Implement the user requirements by selecting appropriate linear and non-linear data structures	Apply
CO6	Modify the existing operations of data structures for changing needs of the software requirements	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	L	L	L	M	-	-	-	L	L	-	L	M	L
CO2	S	M	M	L	M	-	-	-	L	L	-	L	M	M
CO3	S	M	M	L	M	-	-	-	L	L	-	L	M	M
CO4	S	M	M	L	M	-	-	-	L	L	-	L	M	M
CO5	S	M	S	L	M	-	-	-	L	L	-	L	S	S
CO6	S	S	S	L	M	-	-	-	L	L	-	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	5	20																							2	10				
CO2	5		30						50																2	5	10			
CO3													5	20	20					25				2	5	10				
CO4		5	10										5	10										2	5	10				
CO5		5	20						50						20					25				2	5	15				
CO6																20					50							15		
Total (in %)	100						100						100						100											

## Syllabus

**Introduction:** Abstraction - Data structures - Abstract Data Types - Primitive data structures – Types of data structures- Analysis of algorithms - Best, worst and average case time complexities - notations.

**Arrays and Linked Lists:** Data storage and primitive Operations –Singly linked lists, Doubly linked lists, Circular lists - Applications: Addition of Polynomials; Sparse Matrix representation

**Stacks:** Primitive operations - Sequential implementation – Linked Stacks. Applications - Recursion – Expression Processing

**Queues:** Primitive operations - sequential implementation - Dequeues – Applications - Linked queues - Dynamic Storage Management.

**Trees:** Terminologies - Implementation - BINARY TREE: Properties - Sequential and linked representation –Traversals - Expression trees - Infix, Postfix and Prefix expressions - Binary Search Tree -Insertion and deletion – other operations - Need for balancing - AVL trees - Height – searching – insertion and deletion of elements- AVL rotations

**Graph:** Terminologies – Matrix and list representation – Graph Traversals (DFS & BFS)

**Search and Sorting techniques**– Linear Search – Binary search - Bubble sort – Selection sort – Insertion Sort – Quick Sort – Merge Sort

## Reference Books

1. Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, Clifford Stein, "Introduction to Algorithms", Fourth Edition, MIT Press 2022.
2. Sahni Sartaj, "Data Structures, Algorithms and Applications in C++", Silicon Press, 2011.
3. Aaron M. Tanenbaum, Moshe J. Augenstein and Yedidyah Langsam, "Data structures using C and C++", Prentice Hall, 2012.
4. Mark Allen Weiss, "Data structures and Algorithm Analysis in C++", Fourth Edition, 2014.
5. Yashavant Kanetkar, "Data Structures Through C++", 4<sup>th</sup> Edition, BPB Publications, 2022

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>INTRODUCTION</b>	
1.1	Abstraction - Data structures - Abstract Data Types	2
1.2	Primitive data structures – Types of data structures	1
1.3	Analysis of algorithms - Best, worst and average case time complexities	1
1.4	Complexity notations	1
<b>2</b>	<b>Arrays and Linked Lists</b>	
2.1	Data storage and primitive Operations with arrays	2

Module No.	Topic	No. of Periods
2.2	Singly linked lists	2
2.3	Doubly linked lists	2
2.4	Circular linked lists	1
2.5	Addition of Polynomials	1
2.6	Sparse matrix representation	2
<b>3</b>	<b>Stacks</b>	
3.1	Primitive operations - Sequential implementation	2
3.2	Linked Stacks	1
3.3	Recursion	1
3.4	Expression processing	2
<b>4</b>	<b>Queues</b>	
4.1	Primitive operations - sequential implementation	2
4.2	Dequeues- Applications	2
4.3	Linked queues	1
4.4	Dynamic storage management	1
<b>5</b>	<b>Trees</b>	
5.1	Terminologies - Implementation - BINARY TREE: Properties	1
5.2	Sequential and linked representation	1
5.3	Traversals - Expression trees - Infix, Postfix and Prefix expressions	1
5.4	Binary Search Tree -Search and Insertion	2
5.5	Deletion	1
5.6	Applications of BST	1
5.7	Need for balancing - AVL trees - Height – searching – insertion and deletion of elements- AVL rotations	2
<b>6</b>	<b>Graph</b>	
6.1	Terminologies – Matrix and list representation	1
6.2	Graph Traversals (DFS & BFS)	3
<b>7</b>	<b>Search and Sorting techniques</b>	
7.1	Linear Search & Binary search	1
7.2	Bubble sort, Selection sort	1
7.3	Insertion sort	1
7.4	Quick sort	1
7.5	Merge Sort	1
	<b>Total</b>	<b>46</b>

**Course Designer(s):**

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**23DS340 DATABASE MANAGEMENT**

Category	L	T	P	Credit
PCC	4	0	0	4

**Preamble**

This course aims to make it easier for students to comprehend the various DBMS software functionalities and perform numerous operations related to building, modifying, and maintaining databases for real-world applications. It also aims to help students understand various designing concepts, storage solutions, querying methods, and database management.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Formulate Entity Relationship(ER) and Relational Models for a given application to design a consistent database	Apply
CO2	Develop a normalized database for a given application by incorporating various constraints like integrity and value constraints	Apply
CO3	Manipulate relational database using Structured Query Language and relational languages.	Apply
CO4	Apply data design and storage techniques for a given scenario	Apply
CO5	Construct Database with indexes and hash tables for the fast retrieval of data.	Apply
CO6	Illustrate different transaction and concurrency control mechanisms to preserve data consistency in a multi-user environment.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	L	L	L	L	L	-	-	L	L	-	L	M	M
CO2	S	M	M	L	L	L	-	-	L	L	-	L	M	M
CO3	S	M	M	L	L	L	-	-	L	L	-	L	M	M
CO4	S	M	M	L	M	M	-	-	L	L	-	L	M	M
CO5	S	M	S	L	M	S	-	-	L	L	-	L	M	M
CO6	S	S	S	L	M	S	-	-	L	L	-	L	M	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	5	10	15						25																2	5	10			
CO2	5	10	20						25																2	5	15			
CO3		10	25						50					5	15										2	5	15			
CO4														5	15						25				2	10				
CO5													5	10	15						25				2		10			
CO6													5	10		15						50				5		10		
Total (in %)	100						100						100						100											

## Syllabus

**BASIC CONCEPTS:** Introduction to databases – Conventional file Processing – Data Modelling for a database – Three level architecture – Data Independency –Components of a Database Management System (DBMS) – Advantages and disadvantages of a DBMS –System Environment – Users of DBMS – Transaction Management.

**DATA MODELS:** Introduction – Conceptual data modelling – Motivation - Entities, entity types, various types of attributes, relationships, relationship types - E/R Diagram(ERD) notation - Generalization– Aggregation – Conversion of ERD into relational schema – Introduction to Network data model and Hierarchical data model.

**RELATIONAL DATA MODEL:** Introduction – Keys, relational algebra operators: selection, projection, cross product, various types of joins, division, examples, tuple relation calculus, domain relational calculus.

**RELATIONAL DATABASE MANIPULATION:** Structured Query Language (SQL) - Basic data retrieval –nested queries - correlated and uncorrelated - SQL Join – Views.

**DATABASE DESIGN THEORY:** Functional dependencies – Normal forms - Dependency theory – Armstrong's axioms for FDs - Closure of a set of FDs, Minimal covers– 1NF, 2NF, 3NF and BCNF - Join dependencies and definition of 5NF – Examples.

**DATA STORAGE AND INDEXING:** Storage device Characteristics – Operations on file - Sequential files - Index Sequential files – Direct files – Indexing using Tree structures.

**DATABASE SECURITY, INTEGRITY AND CONTROL:** Security and Integrity threats – Defence mechanisms - Transaction processing – concepts - ACID properties - concurrency control - recovery methods.

## Reference Books

1. Silberschatz A., Korth H. and Sudarshan S., “Database System Concepts”, 7<sup>th</sup> edition McGraw Hill, 2021.
2. Elmasri R. and Navathe S.B., “Fundamentals of Database Systems”, Pearson Education, 2017.
3. Raghu Ramakrishnan and Johannes Gehrke, “Database Management System”, McGraw Hill, 2017.
4. Bipin C.Desai, “An Introduction to Database System”, Galgotia Publishers, 2012.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>Basic Concepts</b>	
1.1	Introduction to databases and conventional file processing	1
1.2	Data modelling for a database	2
1.3	Three-level architecture	1
1.4	Data independency	1
1.5	Components of a DBMS, Advantages and disadvantages of a DBMS	1
1.6	System environment & users of DBMS	1
1.7	Transaction management	1
<b>2</b>	<b>Data Models</b>	

Module No.	Topic	No. of Periods
2.1	Introduction	1
2.2	Conceptual data modelling and motivation	2
2.3	Entities, entity types, various types of attributes, relationships, relationship types	2
2.4	E/R Diagram notation	1
2.5	Generalization and aggregation	1
2.6	Conversion of ERD into relational schema	1
<b>3</b>	<b>Relational Data Model</b>	
3.1	Introduction	1
3.2	Keys	1
3.3	Relational algebra operators: selection, projection, cross product, various types of joins, division, examples	2
3.4	Tuple relation calculus, domain relational calculus	2
3.5	Network data model and Hierarchical data model	1
<b>4</b>	<b>Relational Database Manipulation</b>	
4.1	Structured Query Language (SQL) - Basic data retrieval	1
4.2	Nested queries - correlated and uncorrelated	1
4.3	SQL Join – Views	1
<b>5</b>	<b>Database Design Theory</b>	
5.1	Functional dependencies	1
5.2	Normal forms	2
5.3	Dependency theory and Armstrong's axioms for FDs	2
5.4	Closure of a set of FDs, minimal covers	3
5.5	1NF, 2NF, 3NF and BCNF Join dependencies and definition of 5NF	3
<b>6</b>	<b>Data Storage and Indexing</b>	
6.1	Storage device characteristics	1
6.2	Operations on file	1
6.3	Sequential files	1
6.4	Index Sequential files, direct files, indexing using Tree structures	1
<b>7</b>	<b>Database Security, Integrity and Control</b>	
7.1	Security and integrity threats	1
7.2	Defence mechanisms	1
7.3	Transaction processing concepts, ACID properties, concurrency control, recovery methods	2
	<b>Total</b>	<b>46</b>

**Course Designer(s):**

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**23DS350****OPERATING SYSTEMS**

Category	L	T	P	Credit
PCC	3	0	0	3

**Preamble**

The objective of this course is to introduce basic principles of operating systems which include memory management, device management, process management, file management and security & protection mechanisms with case studies from Windows and Linux operating systems.

**Prerequisite**

NIL

**Course Outcomes**

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

CO1	Describe the evolution, types, structure, and functions of operating systems	Understand
CO2	Demonstrate the process management functionalities of operating systems including threads, inter process communication and scheduling	Apply
CO3	Provide solutions for process synchronization and deadlocks	Analyze
CO4	Demonstrate the different memory management policies with paging, segmentation, and virtual memory schemes	Apply
CO5	Determine the mechanisms adopted for I/O and file systems in user applications	Apply
CO6	Execute Linux basic commands and shell scripts	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	L	L	-	L	-	-	-	L	L	-	L	L	L
CO2	S	M	M	L	L	L	-	-	L	L	-	L	L	L
CO3	S	S	S	L	L	L	-	-	L	L	-	M	L	L
CO4	S	M	M	L	L	L	-	-	L	L	-	L	L	L
CO5	S	M	M	L	L	L	-	-	L	L	-	L	L	M
CO6	S	M	S	L	S	L	-	-	L	L	-	M	L	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	10	10																							2	5				
CO2	10	10	20						50																2	5	15			
CO3			20	20						50															2	5		15		
CO4											10	10	20								50				2	5	15			
CO5											10	10	20								25				2	5	10			
CO6													20								25							10		
Total (in %)	100						100						100						100						100					

## Syllabus

**Introduction:** Need of Operating system, Operating system- Operations, functionalities, computing environments – Operating System structures

**Process Management:** Process states — Operations on process – Inter process communication - Thread concepts - Processor Scheduling – Implementation of scheduler - Case studies

**Process Synchronization** - Semaphores, Shared memory multiprocessors, Alternative Synchronization primitives, Monitors, Classical Synchronization problems, Implementation - Bounded Buffer Problem, Reader's Writer's Problem, Dining Philosopher's Problem - Deadlock prevention, avoidance, detection and recovery.

**Memory Management:** Swapping, Paging, Segmentation, Virtual Memory — Demand paging, Page Replacement algorithms – Implementation of algorithms

**File & Storage Management:** File System concepts and access methods, File structure, Allocation methods, free space management, Disk Structure, Disk Scheduling.

**Linux Programming** - Command Line and Shell Scripting Basics.

## Reference Books

1. Abraham Silberschatz, Greg Gagne, Peter B. Galvin, "Operating System Concepts", 9th edition, Wiley, 2013.
2. Andrew Tanenbaum, Herbert Bos, "Modern Operating Systems", Fifth Edition, Pearson Education, 2014
3. William Stallings, "Operating systems Internal and Design Principle", Ninth Edition, Pearson Education, 2015
4. Carl Albing, Vossen, "bash Cookbook: Solutions and Examples for bash Users", 2<sup>nd</sup> Edition, O'Reilly, 2017.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>INTRODUCTION</b>	
1.1	Need of Operating system, Operating system- Operations	2
1.2	Functionalities - computing environments	1
1.3	Operating System structure	1
<b>2</b>	<b>Process Management</b>	
2.1	Process states — Operations on process	1
2.2	Interprocess communication	1
2.3	Thread concepts	1
2.4	Processor Scheduling - implementation	2
<b>3</b>	<b>Process Synchronization.</b>	
3.1	Semaphores	2

Module No.	Topic	No. of Periods
3.2	Shared memory multiprocessors, Alternative Synchronization primitives	1
3.3	Monitors, Classical Synchronization problems,	2
3.4	Implementation - Bounded Buffer Problem, Reader's Writer's Problem, Dining Philosopher's Problem	2
3.5	Deadlock prevention, avoidance, detection and recovery	2
<b>4</b>	<b>Memory Management</b>	
4.1	Swapping techniques	1
4.2	Paging	2
4.3	Segmentation	1
4.4	Virtual Memory — Demand paging	1
4.5	Page Replacement algorithms – Implementation of algorithms	2
<b>5</b>	<b>File &amp; Storage Management</b>	
5.1	File System & access methods	2
5.2	Allocation methods	1
5.3	Free space management	1
5.4	Disk Structure, Disk Scheduling	2
<b>6</b>	<b>Linux Programming</b>	
6.1	Command Line Basics.	2
6.2	Shell Scripting	2
	<b>Total</b>	<b>35</b>

**Course Designer(s):**

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<b>23DS370</b>	<b>DATA STRUCTURES LAB</b>	Category	L	T	P	Credit
		PCC	0	0	4	2

**Preamble**

The objective of this lab is to provide hands-on experience to implement the algorithms for performing various operations on data structures. Students will gain practical knowledge by writing and executing programs in C++ using various data structures such as arrays, linked lists, stacks, queues, trees, graphs and search trees.

**Prerequisite**

- 23DS170 C Programming Lab
- 23DS280 Object Oriented Programming Lab

**Course Outcomes**

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

CO1	Design algorithms for the given problem specifications by choosing appropriate data structures and relevant operations.	Apply
CO2	Implement programs for the given algorithm specification using data structures	Apply
CO3	Implement programs to implement linear data structures: Stack and Queue using arrays and linked list in an application context.	Apply
CO4	Implement non-linear data structures: Graph, Trees in an application context	Apply
CO5	Implement algorithms that use sorting and searching as sub-procedures.	Apply
CO6	Prepare laboratory reports for the applications developed	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1:	S	M	L	-	-	M	-	-	L	-	-	-	M	M
CO2:	S	M	L	-	-	M	-	-	L	-	-	-	M	M
CO3:	S	M	L	-	-	M	-	-	L	-	-	-	M	M
CO4:	S	M	L	-	-	M	-	-	L	-	-	-	M	M
CO5:	S	M	L	-	-	M	-	-	L	-	-	-	M	M
CO6:	S	M	L	-	-	M	-	-	L	-	-	-	M	M

S- Strong; M-Medium; L-Low

### List of Experiments

1. Search algorithms (Linear Search and Binary Search) using Arrays.
2. Sorting algorithms (Bubble/Insertion/Selection and Merge/Quick) using Arrays.
3. Implementation of Singly, Doubly and Circular Linked Lists
4. Stack implementation and Two-way stack implementation
5. Expression evaluation using a Stack
6. Implementation of Queue, Circular Queue
7. Binary Trees and traversals
8. Graph representation and traversal
9. Use of Data Structure libraries in the language

### Reference Books

1. Thomas H. Cormen, Charles E. Leiserson, and Ronald L' Rivest, "Introduction to Algorithms", Fourth Edition, MIT Press, 2022.
2. H.M. Deitel and P.J. Deitel, C "How to program Introducing C++ and Java", Fourth Edition, Pearson Prentice Hall, (2012), Reprint 2020.
3. Reema Thareja "Computer fundamentals and programming in c", Oxford University, Second edition, 2017

### Course Designer(s):

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**23DS380 RELATIONAL DATABASE LAB**

Category	L	T	P	Credit
PCC	0	0	4	2

**Preamble**

This course aims to provide students with considerable knowledge to database design and E-R modelling. This course also provides students with hands-on training in SQL and programming languages that extend SQL.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Model Entity Relationship with E-R diagrams for given application specification.	Apply
CO2	Design database schema considering normalization and relationships within database.	Apply
CO3	Write SQL queries to user specifications and design the queries that are scalable and optimised.	Apply
CO4	Develop triggers, procedures, user defined functions and design PLSQL programs in Databases	Apply
CO5	Use the database for a real world application and connect it from a front end application.	Apply
CO6	Use PL/SQL procedures for a solving a programming logic and develop package for database.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1.	S	S	L	-	L	-	-	-	-	-	-	-	M	L
CO2.	S	S	L	-	L	-	-	-	-	-	-	-	M	L
CO3	S	S	S	-	S	-	-	-	-	-	-	-	M	L
CO4	S	S	S	-	S	-	-	-	-	-	-	-	M	L
CO5.	S	S	S	-	S	-	-	-	-	-	-	-	M	L
CO6	S	M	M	-	S	-	-	-	-	-	-	-	M	L

S- Strong; M-Medium; L-Low

**List of Experiments**

1. Model ER diagram based on given project specifications
2. Design normalized relational database design on given project specification
3. SQL DDL-Create a SQL DDL Queries for a specific application.
4. SQL DML-Insert values into the Tables and Manipulate the data in it using DML commands.(Aggregate functions, Subqueries, Joins, Nested tables, Views)
5. PL/SQL Programming Procedures and functions
6. Implementation of Cursors.
7. Implementation of Triggers.
8. DCL and TCL commands
9. Development of front end application that connects with a database

### Reference Books

1. Silberschatz A., Korth H. and Sudarshan S., "Database System Concepts", McGraw Hill, 2021.
2. Elmasri R. and Navathe S.B., "Fundamentals of Database Systems", Pearson Education, 2017.
3. Raghu Ramakrishnan and Johannes Gehrke, "Database Management System", McGraw Hill, 2017.
4. Bipin C.Desai, "An Introduction to Database System" , Galgotia Publishers, 2012.

### Course Designer(s):

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**23DS410**

**LINEAR ALGEBRA**

Category L T P Credit  
FC 4 0 0 4

**Preamble**

This course will enable students to solve the given system of linear equations through matrices, compute Eigen values, Eigen vectors and model to a quadratic form and construct a singular value decomposition for the given matrix, also perform diagonalization of a given matrix, verify whether the given set is a vector space or not, determine the matrix for the given linear transformation.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1:	Apply the concept of Gauss elimination in solving linear equations and to find the inverse of a matrices	Apply
CO2:	Apply the concept of Eigen values and vectors to orthogonal diagonalizing a given matrix,	Apply
CO3:	Diagonalize a given matrix from quadratic form	Apply
CO4:	Apply the concept of vector spaces, linear independence, basis and dimensions to find Rank	Apply
CO5:	Apply the concept of linear transformation in matrices and geometry linear operators	Apply
CO6:	Apply the concepts of Inner product spaces and orthogonal projections to find QR decomposition and least squares	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO3.	L	L	L	M	-	-	-	-	-	-	-	-	L	M
CO4.	S	S	M	M	-	-	-	-	-	-	-	-	L	S
CO3	S	S	M	M	-	-	-	-	-	-	-	-	L	S
CO4	S	S	M	S	-	-	-	-	-	-	-	-	L	S
CO5	S	S	M	S	-	-	-	-	-	-	-	-	L	S
CO6	S	S	L	L	-	-	-	-	-	-	-	-	L	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	2		20						30																	5	8			
CO2	4	10	20						35																2	5	10			
CO3	4	10	30						35																2		15			
CO4													4	10	30									2	5	12				
CO5													4		20									2		10				
CO6													2	10	20									2	5	15				
Total (in %)	100						100						100						100											

**Syllabus****SYSTEM OF LINEAR EQUATIONS AND MATRICES:**

System of linear equations, Gauss Elimination, Gauss – Jordan, Inverse of a matrix Using Gauss - Jordan.

**EIGEN VALUES AND EIGEN VECTORS:**

Introduction to Eigen values and Eigen Vectors - Diagonalizing a matrix – Similarity and Orthogonal diagonalization – Applications of Eigen value problems – Quadratic forms

**VECTOR SPACES:**

Vector spaces and subspaces — Linear combination, Span, Linear independence and dependence - Basis and Dimension of a vector space - Change of basis.

**LINEAR TRANSFORMATION:**

Introduction to linear transformations — General Linear Transformations – Null space and Range – Rank and nullity – Row space and Column space - Matrices of general linear transformation.

**INNER PRODUCT SPACES:**

Inner product, Length, angle and orthogonality — Orthogonal sets — Orthogonal projections — Inner product spaces — Orthonormal basis: Gram-Schmidt process — QR Decomposition - Best Approximation, Least-squares.

**Reference Books and Web Resources**

1. David C. Lay, "Linear Algebra and its Applications", Pearson Education, 6<sup>th</sup> edition 2021.
2. Howard Anton and Chris Rorres, "Elementary Linear Algebra", Wiley, 12<sup>th</sup> edition 2019.
3. Gilbert Strang, "Linear Algebra and its Applications", Thomson Learning, 5<sup>th</sup> edition, 2016.
4. Steven J. Leon, "Linear Algebra with Applications", Prentice Hall, 10<sup>th</sup> edition, 2021.
5. <https://ilasic.org/>
6. <https://www.khanacademy.org/math/linear-algebra>
7. <https://mathworld.wolfram.com/topics/LinearAlgebra.html>

**Course Contents and Lecture Schedule**

Module No.	Topic	No. of Periods
1	SYSTEM OF LINEAR EQUATIONS AND MATRICES:	
1.1	System of linear equations	1
1.2	Gauss Elimination	2
1.3	Gauss – Jordan	2
1.4	Inverse of a matrix using Gauss - Jordan	2
2	EIGEN VALUES AND EIGEN VECTORS:	
2.1	Introduction to Eigen values and Eigen Vectors	2
2.2	Diagonalizing a matrix	2

Module No.	Topic	No. of Periods
2.3	Similarity and Orthogonal diagonalization	2
2.4	Applications of Eigen value problems	1
2.5	Quadratic forms	2
3	VECTOR SPACES:	
3.1	Vector spaces and subspaces	2
3.2	Linear combination, Span, Linear independence and dependence	3
3.3	Basis and Dimension of a vector space	2
3.4	Change of basis.	2
4	LINEAR TRANSFORMATION:	
4.1	Introduction to linear transformations	1
4.2	General Linear Transformations	1
4.3	Null space and Range, Rank and nullity	2
4.4	Row space and Column space	2
4.5	Matrices of general linear transformation.	2
5	INNER PRODUCT SPACES:	
5.1	Inner product, Length, angle and orthogonality	2
5.2	Orthogonal sets – Orthogonal projections	2
5.3	Inner product spaces	1
5.4	Orthonormal basis: Gram-Schmidt process	2
5.5	QR Decomposition - Best Approximation	2
5.6	Least-squares	2
	<b>Total</b>	<b>44</b>

**Course Designer:**

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**23DS420****PREDICTIVE ANALYTICS**

Category L T P Credit

PCC 4 0 0 4

**Preamble**

This course will enable students to apply specific statistical and regression analysis methods to develop and use various regression and classification predictive models and perform time series analysis with suitable techniques

**Prerequisite**

- 23DS220- Applied Statistics

**Course Outcomes**

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

CO1	Demonstrate simple linear regression models with appropriate validation of the models	Apply
CO2	Develop multiple linear regression models with appropriate validation of the models	Analyze
CO3	Formulate methods to improve linear regression models by addressing different data conflicts	Apply
CO4	Develop logistic regression models with appropriate validation of the models	Analyze
CO5	Demonstrate the application of decision trees in classification and regression	Apply
CO6	Perform time series analysis and forecasting using smoothing, moving average and different auto regressive models	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	L	L	M	-	-	L	L	L	-	L	S	S
CO2	S	S	M	M	M	L	L	L	L	L	-	L	S	S
CO3	S	S	M	S	M	L	L	L	L	L	-	L	S	S
CO4	S	S	M	M	M	L	L	L	L	L	-	L	S	S
CO5	S	M	M	M	M	L	L	L	L	L	-	L	S	S
CO6	S	M	M	M	M	L	L	L	L	L	-	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	5	10	20																						2			15		
CO2	5	10	10	20						50															2	5			10	
CO3		10	10							50															2	5	10			
CO4													5	5	5	20							50		2	5	10	5		
CO5													5	5	25								25		2		10			
CO6														10	20								25			5	10			
Total (in %)	100						100						100						100						100					

**Syllabus**

**SIMPLE LINEAR REGRESSION(SLR):** Correlation analysis, Model building, Estimation and interpretation of coefficients, Validation of SLR model, Coefficient of determination, Significance test, Residual analysis, Outlier analysis, Confidence and Prediction intervals.

**MULTIPLE LINEAR REGRESSION(MLR):** Model building, Estimation and interpretation of coefficients, Validation of MLR model: Adjusted R square, Bias and Variance, Underfitting and overfitting, Crossvalidation, Categorical variables, heteroscedasticity, Multi-collinearity and Variance Inflation Factor, Auto correlation, Outlier analysis, Feature selection and transformations, Regularization

**LOGISTIC AND MULTINOMIAL REGRESSION:** Logistic function, Estimation of probability using Logistic regression, Model validation:Wald Test, Hosmer Lemshow Test, Classification Table and ROC curve, Feature selection, Multinomial Logistic regression

**DECISION TREES :** Introduction,Gini Impurity index and Entropy, Classification and Regression trees

**TIME SERIES ANALYSIS:** Time-series data components, Forecasting techniques, Moving average, Exponential Smoothing: Single, Double and Triple, Regression model for forecasting, Autoregressive Models-AR,MA,ARMA, ARIMA models.

**CASE STUDIES****Reference Books and web resources**

1. Anderson, Sweeney and Williams, "Statistics for business and economics", Cengage Learning, 13<sup>th</sup> Edition, ,2018
2. Trevor Hastie, Robert, Jerome, "The elements of statistical learning: Datamining, Inference and Prediction", 2<sup>nd</sup> Edition, 2017.
3. Eric Seigel, "Predictive Analytics: The Power to Predict Who Will Click, Buy, Lie, or Die", 2016
4. Ronald E. Walpole, Raymond H. Meyers, Sharon L. Meyers, "Probability and Statistics for Engineers and Scientists", Pearson Education, 9<sup>th</sup> Edition, 2022
5. U. Dinesh Kumar, "Business Analytics: The Science of Data-Driven Decision Making", Wiley, 2017
6. Introduction to Predictive Modeling – Coursera

**Course Contents and Lecture Schedule**

Module No.	Topic	No. of Periods
<b>1</b>	<b>SIMPLE LINEAR REGRESSION(SLR):</b>	
1.1	Correlation analysis	1
1.2	Model building, Estimation and interpretation of coefficients,	2
1.3	Coefficient of determination, Significance Testing	2
1.4	Residual analysis	2
1.5	Outlier analysis	1
1.6	Confidence and Prediction intervals	2

Module No.	Topic	No. of Periods
<b>2</b>	<b>MULTIPLE LINEAR REGRESSION</b>	
2.1	Model building, Estimation and interpretation of coefficients	2
2.2	Validation of MLR model: Adjusted R square , Bias and Variance, Underfitting and overfitting, cross validation techniques	2
2.3	Categorical variables, heteroscedasticity, Multi-collinearity and Variance Inflation Factor	2
2.4	Auto correlation	1
2.5	Outlier analysis	1
2.6	Feature selection and transformations, Regularization	2
<b>3</b>	<b>LOGISTIC AND MULTINOMIAL REGRESSION</b>	
3.1	Logistic function, Estimation of probability using Logistic regression	1
3.2	Wald test	2
3.3	Hosmer Lemshow Test	1
3.4	Classification Table and ROC curve	2
3.5	Feature selection, Multinomial Logistic regression	2
<b>4</b>	<b>DECISION TREES</b>	
4.1	Gini Impurity index and Entropy	3
4.2	Classification decision trees	2
4.3	Regression trees	2
<b>5</b>	<b>TIME SERIES ANALYSIS</b>	
5.1	Time-series data components	1
5.2	Forecasting techniques, Moving average,	1
5.3	Exponential Smoothing: Single, Double and Triple	2
5.4	Regression model for forecasting	2
5.5	AR,MA,ARMA, ARIMA models	2
5.6	Case Studies	2
	<b>Total</b>	<b>45</b>

**Course Designer(s):**

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23DS430

**DESIGN AND ANALYSIS OF ALGORITHMS**

Category L T P Credit  
PCC 3 1 0 4

**Preamble**

This course will enable students to solve a given problem using an algorithm. Also, it enables them to mathematically analyze the algorithms for its efficiency and effectiveness for choosing a better solution to real time algorithmic problems.

**Prerequisite**

- 23DS330 Data Structures

**Course Outcomes**

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

CO1	Summarize the relevance of algorithms for computational problems solving and real time applications.	Understand
CO2	Differentiate different algorithmic approaches, techniques and methods.	Understand
CO3	Apply paradigms like divide and conquer, greedy and dynamic programming for a given algorithm.	Apply
CO4	Apply optimization techniques for improving the efficiency of graph-based algorithms to solve complex problems.	Apply
CO5	Analyze the worst-case, best-case and average-case running time of algorithms using asymptotics.	Analyze
CO6	Evaluate polynomial reductions for standard problems with an understanding of the intractable complexity classes like NP- Complete and NP-hard.	Evaluate

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	-	-	-	-	-	-	-	-	-	-	L	L
CO2	S	M	-	-	-	-	-	-	-	-	-	-	L	L
CO3	S	S	-	-	-	-	-	-	-	-	-	-	L	L
CO4	S	S	M	-	-	L	L	-	-	-	-	-	L	L
CO5	S	S	L	-	-	-	-	-	-	-	-	-	L	L
CO6	S	S	L	-	-	-	-	-	-	-	-	-	L	L

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	5	10																							2	10				
CO2	10	10							30																2	10				
CO3		10	30						30																2		20			
CO4									40				5	10	20										2		20			
CO5	5			20									5	10		20							50		2	5		10		
CO6														10	10		10						50			5				10
Total (in %)	100						100						100						100											

## Syllabus

**INTRODUCTION:** Fundamentals of algorithmic problem solving - Methods of specifying an algorithm – proving the correctness – analyzing an algorithm, Asymptotic notations, Recurrences – Master theorem.

**DIVIDE AND CONQUER:** Integer multiplication, matrix multiplication

**GREEDY METHOD:** Minimum cost spanning tree (Kruskal and Prim's algorithms), topological sorting, Huffman codes and data compression.

**DYNAMIC PROGRAMMING:** Principles of dynamic programming – 0/1 knapsack problem, all pairs shortest problem, travelling salesman problem.

**GRAPHS:** Definition – Representations– Network representation, shortest path- Dijkstra's algorithm, Graph search methods (Breadth first and depth first traversals)- Applications of depth first search- Biconnectivity.

**NP AND COMPUTATIONAL INTRACTABILITY:** Basic concepts – Polynomial time reductions, efficient certification and NP, NP hard and NP complete problems.

**COPING WITH NP-COMPLETENESS:** Backtracking-n queen's problem, Graph coloring problem - Branch and bound - 0/1 knap sack problem, traveling salesman problem, Approximation algorithm.

## Reference Books

1. Thomas H. Cormen, Charles E. Leiserson, and Ronald LRivest, —Introduction to Algorithms, Fourth Edition, MIT Press, 2022. (Chapters: 1 to 5, 15, 16, 17, 34 & 35).
2. M.A.Weiss, "Data Structures and Algorithm Analysis in C++", Pearson Education Asia, 2013.
3. Jon Kleinberg and Eve Tardos, "Algorithm Design", Pearson Education, 2012
4. Anany Levitin, "Introduction to Design and Analysis of Algorithms", Pearson Education, 2012.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>Introduction</b>	1
1.1	Fundamentals of algorithmic problem solving	1
1.2	Methods of specifying an algorithm	1
1.3	proving the correctness	1
1.4	analyzing an algorithm	1
1.5	Asymptotic notations	1
1.6	Recurrences	1
1.7	Master theorem	2
2	<b>Divide and Conquer Method (DCM)</b>	
2.1	Introduction to DCM	1

Module No.	Topic	No. of Periods
2.2	Integer Multiplication	1
2.3	Matrix multiplication	1
3	<b>Greedy Method</b>	
3.1	Minimum cost spanning tree (Kruskal and Prim's algorithms)	2
3.2	Topological sorting	1
3.3	Huffman codes	2
3.4	Data compression	1
4	<b>Dynamic programming</b>	
4.1	Principles of Dynamic programming	1
4.2	0/1 knapsack problem	1
4.3	All pairs shortest problem	1
4.4	Travelling salesman problem	1
5	<b>Graphs</b>	
5.1	Definition and Representations	1
5.2	Network representation	1
5.3	Shortest path- Dijkstra's algorithm	2
5.4	Graph search methods (Breadth first and depth first traversals)	2
5.5	Applications of depth first search	1
5.6	Biconnectivity.	1
6	<b>NP and Computational Intractability</b>	
6.1	Basic concepts	1
6.2	Polynomial time reductions	1
6.3	Efficient certification and NP	1
6.4	NP hard problem	1
6.5	NP complete problems	1
7	<b>Coping with NP Completeness</b>	
7.1	Backtracking	1
7.2	N queen's problem	1
7.3	Graph coloring problem	2

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
7.4	branch and bound	1
7.5	0/1 knap sack problem	1
7.6	Travelling salesman problem	2
7.7	Approximation algorithm	1
	<b>Total</b>	<b>44</b>

**Course Designer:**

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**23DS440****ADVANCED DATA  
STRUCTURES**

Category	L	T	P	Credit
PCC	4	0	0	4

**Preamble**

This course aims at facilitating the student to understand the various advanced data structures, their operations and apply them in real world problems.

**Prerequisite**

- 23DS330 Data Structures

**Course Outcomes**

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

CO1	Demonstrate the operations of Balanced Binary search trees and related applications	Apply
CO2	Implement algorithms for Multiway Search Trees	Apply
CO3	Explain the concept of Multi-dimensional Search Trees	Understand
CO4	Implement Priority Queues with Binary heaps	Apply
CO5	Show the avoidance of collisions in the hash tables using collision resolution techniques including open and closed hashing techniques.	Apply
CO6	Explain the implementation of priority queues with leftist heaps and binomial heap	Understand
CO7	Implement suitable technique and algorithms for disjoint set operations	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	L	L	L	M	-	-	L	L	-	L	M	L
CO2	S	M	L	L	L	M	-	-	L	L	-	L	M	L
CO3	S	-	L	L	-	-	-	-	-	-	-	-	M	L
CO4	S	M	L	L	L	M	-	-	L	L	-	M	M	L
CO5	S	M	L	L	L	M	-	-	L	L	-	M	M	L
CO6	S	-	L	L	-	-	-	-	-	L	-	M	M	L
CO7	S	M	L	L	L	M	-	-	L	L	-	L	M	L

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	10		20						50																			15		
CO2	5	20	20						50																5			10		
CO3	5	20																							5	5				
CO4													5	10	20													15		
CO5													5		20												5	15		
CO6													5	10											5	5				
CO7														5	20													15		
Total (in %)	100						100						100						100											

**Syllabus****Balanced Binary Search Trees:** – Binary Search Tree - Red Black Tree –Splay Tree**Hash Table:** Hash function – Separate chaining – Open addressing – Linear probing – Quadratic probing – Double hashing – rehashing**Priority Queues (Heaps):** Binary Heaps- Leftist Heaps - Property and operations- Binomial heap, Fibonacci Heap**Multway Search Trees:** B-Tree, - B+ trees, B\*-trees, Tries, Application on Tries: Pattern Searching, String matching, Digital search trees, Dictionary applications**Multidimensional Search Trees:** Range search–k-d trees- Quad trees.**Disjoint Sets:** Disjoint set operations-linked list representation of disjoint sets, disjoint set forests, tree representation, union by rank – Applications of disjoint sets**Reference Books**

1. Thomas H. Cormen, Charles E. Leiserson and Ronald L. Rivest, "Introduction to Algorithms", Fourth Edition, MIT Press, 2022.
2. Karumanchi Narasimha, "Data Structures and Algorithms Made Easy", 2016
3. Devraj Ganguly, "Introduction To Data Structures And Algorithms: A Conceptual Guide", 2021

**Course Contents and Lecture Schedule**

Module No.	Topic	No. of Periods
<b>1</b>	<b>Balanced Binary Search Trees</b>	
1.1	Binary Search Tree	1
1.2	Red Black Tree	2
1.3	Splay Tree	2
	Practice Problems	2
<b>2</b>	<b>Hash Table:</b>	
2.1	Hash function	1
2.2	Separate chaining	1
2.3	Open addressing	1
2.4	Linear probing, Quadratic probing	1
2.5	Double hashing	2
2.6	Rehashing	1
	Practice Problems	2
<b>3</b>	<b>Priority Queues (Heaps):</b>	

Module No.	Topic	No. of Periods
3.1	Binary Heaps – property and operations	1
3.2	Leftist Heaps- property and operations	2
3.3	Binomial heap – property and operations	2
3.4	Fibonacci Heap – property and operations	2
	Practice Problems	2
<b>4</b>	<b>Multway Search Trees:</b>	
4.1	B-Tree	1
4.2	B+ Tree	1
4.3	B* Tree	1
4.4	Tries	1
4.5	Application on Tries:,	1
4.6	Pattern Searching and String Matching	1
4.7	Digital search trees and Dictionary applications	1
	Practice Problems	2
<b>5</b>	<b>Multidimensional Search Trees:</b>	
5.1	k-d trees	1
5.2	Quad trees	1
	Practice Problems	2
<b>6</b>	<b>Disjoint Sets:</b>	
6.1	Disjoint set operations	1
6.2	Linked list representation of disjoint sets	1
6.3	Disjoint set forests -tree representation	1
6.4	Union by rank, Applications of Disjoint sets	1
	Practice Problems	2
	<b>Total</b>	<b>44</b>

**Course Designer(s):**

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**23DS450 SOFTWARE ENGINEERING**

Category L T P Credit

PCC 3 0 0 3

**Preamble**

This course will provide knowledge of software engineering discipline, to analyze risk in software design and quality, to introduce the concept of advance software methodology and to plan, design, develop and validate the software project.

**Prerequisite**

NIL

**Course Outcomes**

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

CO1: Identify appropriate process models in the software industry according to given circumstances	Understand
CO2: Demonstrate the concepts of DevOps and mastering aspects of software development and automated building.	Understand
CO3: Apply Project Management and Requirement analysis principles to S/W project development.	Apply
CO4: Develop Design models in software engineering effectively, the architecture, the user interface, and the component level.	Apply
CO5: Analyze the cost estimate and problem complexity using various estimation techniques	Analyze
CO6: Generate test cases using the techniques involved in selecting: (a) White Box testing (b) Block Box testing	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	L	-	-	-	-	-	-	-	-	-	-	M	L
CO2	M	L	-	-	-	M	-	-	-	-	-	-		L
CO3	S	M	L	-	-	-	-	S	-	-	-	-	L	L
CO4	S	M	L	-	-	-	-	-	S	-	-	-	L	
CO5	S	S	M	-	-	-	-	-	-	-	-	-	L	
CO6	S	S	M	-	-	-	-	-	-	M	-	-	L	L

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	10	10							20																		10			
CO2	10	10							20																		10			
CO3			60						60																		10	20		
CO4												10			20												20			
CO5														10		30										5		10		
CO6															30											5		10		
Total (in %)	100						100						100						100											

## Syllabus

**Software engineering** - Process framework - Umbrella activities. Software process models - Waterfall - Prototyping - evolutionary process model - Agile model - Scrum Teams and Artifacts - Sprint Planning Meeting - Daily Scrum Meeting - Sprint Review Meeting - DevOps.

**Requirements Engineering** - Inception - Elicitation - Elaboration - Negotiation - Specification - Validation. Requirements Management - Identifying Stakeholders - Non-functional Requirements - Traceability. Requirements Gathering - Collaborative Requirements Gathering - Usage Scenarios - Elicitation Work Products - Developing Use Cases -Building the Analysis Model - Elements of the Analysis Model - Validating Requirements.

**Requirements modelling** - Business Use Case, Class-Based Modelling - Functional Modelling - Behavioral Modelling. Design concepts

**Design model** - Software Architecture - User Experience Design - Mobile App design. Software costs estimation. Software Quality Assurance.

**Software testing** - Black Box - White Box - Unit - Integration - System - Validation.

## Reference Books

1. Roger Pressman, Software Engineering: A Practitioners Approach, (9<sup>th</sup> Edition), McGraw Hill, 2020. Chapters 1,2,3, 7, 8, 9, 10, 12, 13, 17, 19, 20, 25.
2. Ian Somerville, Software Engineering, 10th edition, Pearson, 2021.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>Software</b>	
1.1	Software Engineering- Process framework	1
1.2	Umbrella activities.	1
1.3	Software process models	1
<b>2</b>	<b>Life Cycle Models</b>	
2.1	Water fall, incremental, spiral,	2
2.2	WINWIN spiral,	1
2.3	evolutionary, prototyping	2
2.4	Agile Process Models, Scrum Teams and Artifacts - Sprint Planning Meeting - Daily Scrum Meeting - Sprint Review Meeting	2
2.5	Devops	1
<b>3</b>	<b>Requirements Engineering tasks</b>	

Module No.	Topic	No. of Periods
3.1	Inception - Elicitation - Elaboration	2
3.2	Negotiation - Specification - Validation	2
3.3	Requirements Management - Identifying Stakeholders -	2
3.4	Non-functional Requirements - Traceability.	2
3.5	Collaborative Requirements Gathering	1
3.6	Usage Scenarios - Elicitation Work Products	1
3.7	Developing Use Cases -Building the Analysis Model	1
3.8	Elements of the Analysis Model - Validating Requirements.	2
<b>4</b>	<b>Requirements modelling</b>	
4.1	Business Use Case	1
4.2	Class-Based Modelling	1
4.3	Functional Modelling	1
4.4	Behavioral Modelling	1
<b>5</b>	<b>Design concepts</b>	
5.1	Software Architecture	1
5.2	User Experience Design	1
5.3	Software costs estimation	1
5.4	Software Quality Assurance	1
<b>6</b>	<b>Software testing</b>	
6.1	Black Box - White Box	1
6.2	Unit - Integration -	1
6.3	System-Validation.	1
	Total	35

**Course Designer(s):**

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**23DS470 PREDICTIVE ANALYTICS LAB**

Category	L	T	P	Credit
PCC	0	0	4	2

**Preamble**

This course will provide a practical application of specific statistical and regression analysis methods to develop and use various regression and classification predictive models and perform time series analysis with suitable techniques based on the real time requirements

**Prerequisite**

- 23DS270 Python Programming And Applied Statistics Lab

**Course Outcomes**

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

CO1	Implement linear regression models with appropriate data collection, pre-processing and postprocessing techniques	Apply
CO2	Implement logistic regression models with appropriate data collection, pre-processing and postprocessing techniques	Apply
CO3	Implement decision trees algorithms for classification and regression	Apply
CO4	Perform time series analysis and forecasting using different auto regressive models	Apply
CO5	Design a project or research presentation with an identified societal problem in a team	Create
CO6	Write technical reports for the tasks performed	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO2	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO3	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO4	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO5	S	S	S	S	S	S	S	S	S	S	S	S	S	S
CO6	-	-	-	-	L	L	L	M	S	L	-	L	S	S

S- Strong; M-Medium; L-Low

**List of Experiments**

Experiments shall be done on data available on Universal repositories / self-collected survey data in Python/R

- Implement and validate Simple Linear regression after appropriate data collection and pre-processing
- Implement and validate Multi Linear regression after appropriate data collection and pre-processing with appropriate testing for multicollinearity (with categorical variables)

- Implement and validate Logistic regression
- Implement cross validation in the results obtained for Linear and logistic regression
- Implementing improved performance measures in the results obtained for Linear and logistic regression (Features subset selection, Regularization)
- Implementing Decision trees for classification and regression
- Time Series analysis with moving average and smoothing methods
- Time Series analysis with ARIMA
- Mini project / Research work presentation processing real world user requirements/ research requirements and implementing the same.

#### Reference Books and web resources

1. Wes McKinney, "Python for Data Analysis", Oreilly, 3<sup>rd</sup> Edition, 2022
2. Alvaro Fuentes, "Hands-On Predictive Analytics with Python: Master the complete predictive analytics process, from problem definition to model deployment", Packt, 2018
3. Basic Data Processing and Visualization – Coursera
4. Data Analysis with R Specialization – Coursera

#### Course Designer(s):

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**23DS480****JAVA PROGRAMMING LAB**

Category	L	T	P	Credit
PCC	0	0	6	3

**Preamble**

This course will help the students how to use the Java programming language constructs to create solutions for problems in the real world and develop Java programs for user defined specifications.

**Prerequisite**

- 23DS280 Object Oriented Programming Lab

**Course Outcomes**

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

CO1	Develop Java programs using OOP Principles and proper program structuring.	Apply
CO2	Develop Java programs to implement exception handling and Multithreading applications.	Apply
CO3	Implement Java programs using packages, Streams, interfaces and collection classes	Apply
CO4	Use API that allows Java program to access the database management systems.	Apply
CO5	Implement Java Programs for Events and interactivity using Layout Manager.	Apply
CO6	Create Java programs for network chatting and remote method development to construct an internet application.	Apply
CO7	Implement server side and client-side application using java servlet and java scripting language.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1.	S	M	S	-	L	-	-	-	-	-	-	-	M	L
CO2.	S	S	S	L	L	-	-	-	-	-	-	-	S	M
CO3	S	S	S	M	M	L	-	-	L	-	-	-	S	M
CO4	S	S	S	M	M	L	-	L	M	-	L	-	S	M
CO5	S	S	S	M	M	-	-	-	M	-	-	-	S	M
CO6	S	S	S	M	M	L	-	L	M	-	M	L	S	M
CO7	S	S	S	M	M	L	-	L	M	-	M	L	S	M

S- Strong; M-Medium; L-Low

### List of Experiments

Develop Java Programs for

1. Use of Objects
2. Using classes and inheritance
3. JNI concepts
4. Multithread applications
5. Exception handling
6. Implementing packages and interfaces
7. Streams
8. Java Collection and utility Classes
9. JDBC Connectivity using different statements
10. Applet program for Animation text, images and sounds
11. Events and interactivity using AWT and Java swing.
12. Socket program for network chatting using Inter Process communication.
13. Client server application using RMI techniques
14. Java Servlets
15. Simple front-end design with validation script

### Mini Project in Java

#### Reference Books:

1. Herbert Schildt, "Java: The Complete Reference-12<sup>th</sup> Edition", McGraw Hill, 2021.
2. Paul Deitel and Harvey Deitel, "Java How to Program (Early Objects)", Pearson, Eleventh Edition, 2017.

#### Course Designer(s):

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**23DS510****NUMERICAL METHODS**

Category L T P Credit

FC 4 0 0 4

**Preamble**

The students will be able to apply appropriate methods to solve algebraic, transcendental equations, polynomial equations, simultaneous linear equations and the inverse of a matrix. Also can calculate the solutions for problems related to interpolation, differentiation and integration and apply various methods for finding solutions of Ordinary and Partial Differential Equations.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Solve the algebraic, transcendental and polynomial equations to find the numerical solutions	Apply
CO2	Solve the simultaneous linear equations to find the solutions and also to find inverse of the given matrix	Apply
CO3	Solve the given matrices by applying the concept of Power method and Jacobi method to find the Eigen values and Eigen vectors	Apply
CO4	Apply finite differences concepts and various interpolation methods for finding intermediate values of well-known data	Apply
CO5	Utilize various numerical methods in numerical differentiation and numerical integration related problems in real life statistical data and interpolate the value	Apply
CO6	Solve the ordinary differential equations and find the numerical solutions	Apply
CO7	Obtain the numerical solutions of partial differential equations	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	L	M	L	L	-	-	-	-	L	L	S	M
CO2	S	S	M	M	L	L	-	-	-	-	L	L	S	M
CO3	S	S	S	M	L	L	L	-	-	-	L	L	S	S
CO4	S	S	S	M	M	L	L	L	L	-	L	L	S	M
CO5	S	S	M	S	M	M	L	-	-	-	M	M	S	S
CO6	S	S	S	S	S	M	M	L	L	-	M	M	S	M
CO7	S	S	S	S	S	M	M	L	L	-	M	M	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	4	10	10						20																2	10				
CO2	2	5	20						20																2	10				
CO3	2		20						30																		14			
CO4	2	5	20						30																2		14			
CO5													4	10	20										30					
CO6													4	10	20										40					
CO7													2		30										30					
Total (in %)	100						100						100						100											

## Syllabus

**NUMERICAL SOLUTION OF ALGEBRAIC EQUATIONS:** Types of error, Fixed point iteration method, Bisection method, Method of False position, Newton Raphson method, Bairstow's method.

**NUMERICAL SOLUTION OF ALGEBRAIC SIMULTANEOUS EQUATIONS:** Solution of linear system of equations: Gauss - elimination, Gauss - Jordan, Cholesky method, Crout's method, Gauss – Jacobi method, Gauss – Seidel method, Matrix Inverse by Gauss – Jordan method.

**EIGENVALUES AND EIGENVECTORS:** Power method for finding dominant Eigen value and Eigen vectors, Inverse power method for finding smallest Eigen value, Jacobi method for symmetric matrices for finding dominant Eigen value and Eigen vectors.

**INTERPOLATION AND CURVE FITTING:** Finite difference operators - Interpolating Polynomials, Lagrange's interpolation, Divided Differences, Newton Forward and Backward Differences, Spline Curves methods.

**NUMERICAL DIFFERENTIATION AND INTEGRATION:** Numerical Differentiation - Newton forward, backward and divided differences. Numerical Integration - Trapezoidal rule, Simpson's (1/3)rd and (3/8)th rule, Double Integration - Trapezoidal rule and Simpson's (1/3)rd rule, Gaussian Quadrature.

**NUMERICAL SOLUTION OF ORDINARY DIFFERENTIAL EQUATIONS:** Single step methods: Taylor's series method, Euler and Improved Euler methods, Fourth order Runge – Kutta method, Multistep methods: Milne's predictor - corrector method.

**NUMERICAL SOLUTION OF PARTIAL DIFFERENTIAL EQUATIONS:** Classification of partial differential equations of second order. Solution of two dimensional Laplace's equations by Liebmann's iterative process and Poisson's equations, Solution of one dimensional heat equation using Bender Schmidt and Crank Nicholson difference schemes, Solution of one dimensional wave equation by explicit scheme.

## Reference Books & web resources

1. Steven C. Chapra and Raymond P. Canale, "Numerical Methods for Engineers with Software and Programming Applications", 8th edition, published by McGraw Hill in 2020.
2. Timothy Sauer, Tim Sauer, "Numerical Analysis", 2nd Edition, Published by Pearson, 2014.
3. B.S. Grewal, "Numerical Methods in Engineering and Science", Khanna Publishers, 42nd edition, 2017
4. Veerarajan, T., & Ramachandran, T., "Numerical methods with programs in C", Tata McGraw-Hill, 2014.
5. Sastry, S. S., "Engineering Mathematics" Vol. PHI Learning Pvt. Ltd., 2008.
6. <https://www.math.hkust.edu.hk/~machas/numerical-methods.pdf>
- 7.. <https://www-users.cse.umn.edu/~arnold/8445.f11/notes.pdf>
8. <https://archive.nptel.ac.in/courses/111/107/111107063/>

**Course Contents and Lecture Schedule**

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>1</b>	<b>NUMERICAL SOLUTION OF ALGEBRAIC EQUATIONS</b>	
1.1	Types of error, Fixed point iteration method	2
1.2	Bisection method	1
1.3	Method of False position	1
1.4	Newton Raphson method	1
1.5	Bairstow's method	1
<b>2</b>	<b>NUMERICAL SOLUTION OF ALGEBRAIC SIMULTANEOUS EQUATIONS</b>	
2.1	Solution of linear system of equations: Gauss - elimination, Gauss - Jordan	2
2.2	Cholesky method, Crout's method	2
2.3	Gauss – Jacobi method, Gauss – Seidel method	2
2.4	Matrix Inverse by Gauss – Jordan method	1
<b>3</b>	<b>EIGENVALUES AND EIGENVECTORS</b>	
3.1	Power method for finding dominant Eigen value and Eigen vectors, Inverse power method for finding smallest Eigen value	2
3.2	Jacobi method for symmetric matrices for finding dominant Eigen value and Eigen vectors	2
<b>4</b>	<b>INTERPOLATION AND CURVE FITTING</b>	
4.1	Finite difference operators, Interpolating Polynomials	1
4.2	Lagrange's interpolation	1
4.3	Divided Differences	1
4.4	Newton Forward and Backward Differences	2
4.5	Spline Curves methods	2
<b>5</b>	<b>NUMERICAL DIFFERENTIATION AND INTEGRATION:</b>	
5.1	Numerical Differentiation - Newton forward, backward and divided differences	3
5.2	Numerical Integration - Trapezoidal rule, Simpson's (1/3)rd and (3/8)th rule	2
5.3	Double Integration - Trapezoidal rule and Simpson's (1/3)rd rule	2
5.4	Gaussian Quadrature	2

Module No.	Topic	No. of Periods
<b>6</b>	<b>NUMERICAL SOLUTION OF ORDINARY DIFFERENTIAL EQUATIONS</b>	
6.1	Single step methods: Taylor's series method	2
6.2	Euler and Improved Euler methods	2
6.3	Fourth order Runge – Kutta method	2
6.4	Multistep methods: Milne's predictor - corrector method	2
<b>7</b>	<b>NUMERICAL SOLUTION OF PARTIAL DIFFERENTIAL EQUATIONS</b>	
7.1	Classification of partial differential equations of second order	1
7.2	Solution of two dimensional Laplace's equations by Liebmann's iterative process and Poisson's equations	3
7.3	Solution of one dimensional heat equation using Bender Schmidt and Crank Nicholson difference schemes	2
7.4	Solution of one dimensional wave equation by explicit scheme	1
	<b>Total</b>	<b>48</b>

**Course Designer(s):**

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**23DS520 OPTIMIZATION TECHNIQUES**

Category L T P Credit  
FC 4 0 0 4

**Preamble**

This course will enable students to learn and apply optimization techniques to enhance Linear Programming Problem, Non-Linear Programming Problem, Integer Programming Problem and Dynamic Problems for solving complex real-world problems efficiently. Linear programming for resource allocation, NLPP allows to apply optimization techniques in Quadratic Programming Problem, Dynamic Programming to solve complex problems in Networks and Global optimization finds the absolute best solution.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Apply linear programming techniques like Graphical, Simplex, Big M, and Two Phase & Dual Simplex Methods for optimum solutions.	Apply
CO2	Apply the transportation model for testing the closeness of their results to optimal results. Solve the assignment problem through Hungarian algorithm to domain specific situations.	Apply
CO3	Determine the optimum values of integer programming problems using Gomary's cutting plane method.	Apply
CO4	Apply the concepts of PERT and CPM for decision making and optimally managing projects.	Apply
CO5	Analyse and apply appropriate inventory techniques in domain specific situations.	Apply
CO6	Analyse and Apply the concept of Convex Optimization in Quadratic Programming.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	M	M	S	L	L	M	L	M	M	M	S	M
CO2	M	M	S	M	S	L	L	L	-	M	M	M	M	S
CO3	S	S	S	M	M	-	-	-	-	L	L	L	S	M
CO4	M	M	M	S	M	L	L	M	-	-	L	L	M	M
CO5	M	M	M	S	M	M	M	M	L	M	L	L	M	S
CO6	S	S	S	S	S	M	L	M	L	M	-	-	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	4	10	20				10	20																	2		20			
CO2	4		30				20	20																	2	5	10			
CO3	2	10	20				10	20																	2	10				
CO4													4	10	20									2	5	10				
CO5													2		30									20		20				
CO6													4	10	20									10	20					
Total (in %)	100						100						100						100											

**Syllabus**

**LINEAR PROGRAMMING PROBLEM:** Modelling with linear programming model - Graphical solution for two dimensional problems - Simplex method, Big M- method - Two phase simplex methods. Dual Simplex Method - Special cases in the simplex method.

**TRANSPORTATION & ASSIGNMENT MODEL:** North West Corner Rule, Least Cost Method, and Vogel's Approximation Methods - Test for optimality - Real-Life Application - Hungarian algorithm Variants of the Assignment problem - Travelling Salesman Problem.

**INTEGER LINEAR PROGRAMMING PROBLEM:** Gomory's cutting plane method - Branch and Bound method.

**NETWORK ANALYSIS:** Introduction - Phases of Project Management IN Network Analysis - Critical path Method - Project Evaluation and Review Techniques.

**INVENTORY MODEL** - Inventory Problem: A Supply Chain Perspective - Static EOQ Models - Dynamic EOQ Models – Single & Multi Period Model.

**NON-LINEAR PROGRAMMING PROBLEM:** Convex sets - Convex functions - Convex optimization problems- Direct search Methods – Gradient Method - Quadratic Programming.

**Reference Books & web resources**

1. Hamdy A. Taha: "Operations Research – An Introduction" Prentice Hall of India Pvt Ltd., 10th Edition, 2019.
2. Sharma J.K.: "Operations Research Theory and applications", Macmillan India Ltd., V Edition, 2015.
3. Chandrasekara Rao, K. Shanti Lata Misra, "Operation Research", Alpha science international Ltd-2015.
4. Kanti Swarup, P.K.Gupta and Man Mohan, "Operations Research", Sultan Chand, 20<sup>th</sup> Edition, 2022.
5. <https://www.khanacademy.org/math/multivariable-calculus/applications-of-multivariable-derivatives/constrained-optimization/a/lagrange-multipliers-examples>
6. <https://archive.nptel.ac.in/courses/111/102/111102012/>

**Course Contents and Lecture Schedule**

Module No.	Topic	No. of Periods
<b>1</b>	<b>LINEAR PROGRAMMING</b>	
1.1	Modelling with linear programming model	1
1.2	Graphical solution for two dimensional problems	1
1.3	Simplex method, Big M- method	1
1.4	Two phase simplex methods	2
1.5	Dual Simplex Method	2
1.6	Special cases in the simplex method	1
<b>2</b>	<b>TRANSPORTATION &amp; ASSIGNMENT MODEL</b>	

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
2.1	Introduction, Mathematical Model of Transportation Problem	1
2.2	Methods for Finding Initial Solution using NWCR & LCM	2
2.3	Method for Finding Initial Solution using VAM	1
2.4	Test for Optimality	1
2.5	Mathematical Models of Assignment Problem	1
2.6	Hungarian Method for Solving Assignment Problem	1
2.7	Variations of the Assignment Problem	1
<b>3</b>	<b>INTEGER LINEAR PROGRAMMING</b>	
3.1	Introduction - Types of Integer Programming Problems	1
3.2	Gomory's cutting plane method	2
3.3	Gomory's Mixed-Integer Cutting Plane Method	2
3.4	Branch and Bound Method	2
<b>4</b>	<b>NETWORK ANALYSIS</b>	
4.1	Introduction, Basic Differences Between PERT and CPM	2
4.2	Phases of Project Management	2
4.3	Critical Path Analysis	2
4.4	Project Evaluation and Review Techniques	2
<b>5</b>	<b>INVENTORY CONTROL</b>	
5.1	Inventory Model - A Supply Chain Perspective	1
5.2	Static EOQ Models	2
5.3	Dynamic EOQ Models	2
5.4	Single & Multi Period Model	2
<b>6</b>	<b>NON-LINEAR PROGRAMMING</b>	
6.1	Convex sets	1
6.2	Convex functions	2
6.3	Convex optimization problems	1
6.4	Direct search Methods	1
6.5	Gradient Method	1

Module No.	Topic	No. of Periods
6.6	Quadratic Programming Problem	2
	<b>Total</b>	<b>46</b>

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**23DS530****WEB TECHNOLOGY**

Category L T P Credit

PCC 4 0 0 4

**Preamble**

This syllabus on Web Technology introduces essential concepts and practices in web development, covering both client-side and server-side programming. It emphasizes key technologies such as HTML, CSS, JavaScript, PHP, JSP, and database access. Students will develop practical skills to create dynamic and interactive web applications.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Understand the fundamental concepts of web technology, including the structure and function of web applications.	Understand
CO2	Create web pages using HTML, CSS, and JavaScript for interactive user experiences.	Apply
CO3	Develop server-side applications using PHP and handle web form data efficiently.	Apply
CO4	Use XML and related technologies for data interchange and presentation in web applications.	Apply
CO5	Deploy and manage web applications using web servers and servlets, ensuring proper session handling and data management.	Apply
CO6	Implement databases connectivity using JDBC and PHP for a web applications	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	S	S	-	-	M	-	-	-	-	-	S	S
CO2	S	S	M	S	L	-	M	-	S	-	-	-	S	S
CO3	M	M	S	M	-	-	-	-	-	-	M	-	S	M
CO4	S	S	S	S	M	-	-	-	-	M	-	-	S	M
CO5	L	S	M	M	S	S	-	-	M	-	S	-	S	S
CO6	S	S	S	S	M	L	-	M	M	-	-	S	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	10	10					20																		10	10				
CO2	10	10	20						40																	10	10			
CO3		10	30						40																		20			
CO4													10	10	10					10	10						10			
CO5													10	10	30					40						10	10			
CO6															20						40						10			
Total (in %)	100						100						100						100											

## Syllabus

**WEB BASICS AND HTML: Introduction to Internet:** World Wide Web, Web Browsers, URL, MIME, HTTP, Web Programmer's Tool Box – **HTML Common Tags:** Lists, Tables, Images, Forms, Frames.

**CSS AND CLIENT-SIDE PROGRAMMING: Basics of CSS:** Types of CSS - **Client-Side Programming (JavaScript):** Introduction to JavaScript, Declaring Variables, Functions, Event Handlers (onclick, onsubmit, etc.), Form Validation.

**SERVER-SIDE PROGRAMMING WITH PHP: PHP Basics:** Declaring Variables, Data Types, Operators, Control Structures, Functions - **Web Form Data Handling:** Reading data from web form controls like text, buttons, radio buttons, lists, etc.

**ADVANCED PHP AND XML: Advanced PHP:** Handling File Uploads, Sessions, and Cookies - **Introduction to XML:** Document Type Definition (DTD), XML Schemas, Document Object Model (DOM), Presenting XML.

**WEB SERVERS, SERVLETS, AND JSP: Web Servers:** Tomcat Web Server, Installing the Java Software Development Kit, Tomcat Server & Testing Tomcat - **Introduction to Servlets:** Servlet Lifecycle, JSDK, Deploying Servlet, The Servlet API, The javax.servlet Package - **JSP Overview:** The Problem with Servlets, Anatomy of a JSP Page, JSP Processing.

**DATABASE ACCESS AND JAVABEANS: Database Programming using JDBC:** JDBC Drivers, Studying javax.sql. \* Package, Accessing a Database from a Servlet and JSP Page - **PHP Database Connectivity:** Connecting to a Database in PHP, Executing Simple Queries - **JavaBeans:** Introduction to JavaBeans, Deploying JavaBeans in a JSP Page.

## Reference Books

1. Uttam K. Roy, "Web Technologies", Oxford University Press, 1st Edition (2010).
2. Steven Holzner, "PHP: The Complete Reference", McGraw-Hill, 1st Edition (2008).
3. Chris Bates, "Web Programming: Building Internet Applications", 3rd Edition, Wiley (2010).
4. Hans Bergsten, "Java Server Pages", O'Reilly, 3rd Edition (2003).
5. Jon Duckett, "Beginning Web Programming", WROX, 2nd Edition (2012).
6. R.W. Sebesta, "Programming the World Wide Web", 8th Edition, Pearson (2015).
7. Deitel & Deitel, "Internet and World Wide Web: How to Program", 6th Edition, Pearson (2011).
8. Paul S. Wang, "An Introduction to Web Design and Programming", 1st Edition, Cengage Learning (2003).
9. Marty Hall & Larry Brown, "Core Servlets and JavaServer Pages: Core Technologies", Volume 1, 2nd Edition, Pearson (2003).

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>WEB BASICS AND HTML</b>	
1.1	Introduction to Internet: World Wide Web, Web Browsers, URL, MIME, HTTP, Web Programmer's Tool Box.	4

1.2	HTML Common Tags: Lists, Tables, Images, Forms, Frames.	4
<b>2</b>	<b>CSS AND CLIENT-SIDE PROGRAMMING</b>	
2.1	Basics of CSS: Types of CSS.	3
2.2	Client-Side Programming (JavaScript): Introduction to JavaScript, Declaring Variables, Functions, Event Handlers (onclick, onsubmit, etc.), Form Validation.	5
<b>3</b>	<b>SERVER-SIDE PROGRAMMING WITH PHP</b>	
3.1	PHP Basics: Declaring Variables, Data Types, Operators, Control Structures, Functions.	5
3.2	Web Form Data Handling: Reading data from web form controls like text, buttons, radio buttons, lists, etc.	4
<b>4</b>	<b>ADVANCED PHP AND XML</b>	
4.1	Advanced PHP: Handling File Uploads, Sessions, and Cookies	5
4.2	Introduction to XML: Document Type Definition (DTD), XML Schemas, Document Object Model (DOM), Presenting XML.	4
<b>5</b>	<b>WEB SERVERS, SERVLETS, AND JSP</b>	
5.1	Web Servers: Tomcat Web Server, Installing the Java Software Development Kit, Tomcat Server & Testing Tomcat.	3
5.2	Introduction to Servlets: Servlet Lifecycle, JSDK, Deploying Servlet, The Servlet API, The javax.servlet Package.	4
5.3	JSP Overview: The Problem with Servlets, Anatomy of a JSP Page, JSP Processing.	4
<b>6</b>	<b>DATABASE ACCESS AND JAVABEANS</b>	
6.1	Database Programming using JDBC: JDBC Drivers, Studying Javax.sql.* Package, Accessing a Database from a Servlet and JSP Page.	4
6.2	PHP Database Connectivity: Connecting to a Database in PHP, Executing Simple Queries.	3
6.3	JavaBeans: Introduction to JavaBeans, Deploying JavaBeans in a JSP Page.	3
	<b>Total</b>	<b>44</b>

**Course Designer:**

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**23DS540****MACHINE LEARNING**

Category L T P Credit

PCC 4 0 0 4

**Preamble**

The course aims to provide an theoretical foundations and hands-on applications of Machine Learning and impart advanced knowledge on the foundational principles of Predictive Analytics

**Prerequisite**

- 23DS220- Applied Statistics
- 23DS420 – Predictive Analytics

**Course Outcomes**

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

CO1	Apply various parametric methods like Maximum Likelihood Estimation and Bayes' Estimator for classification and regression tasks with a knowledge of Bayesian decision theory	Apply
CO2	Identify methods to handle multivariate data on performing classification and regression using multivariate techniques including parametric methods	Analyze
CO3	Apply dimensionality reduction techniques of subset selection, PCA and LDA to simplify high-dimensional data problems	Apply
CO4	Apply non-parametric methods to perform classification, regression, and clustering on multivariate data	Apply
CO5	Implement gradient descent and logistic discrimination algorithms for classification problems.	Apply
CO6	Construct kernel methods for implementing maximum margin classifiers	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	M	M	M	L	L	L	L	L	-	L	S	S
CO2	S	S	M	M	M	L	L	L	L	L	-	L	S	S
CO3	S	M	M	M	M	L	L	L	L	L	-	L	S	S
CO4	S	M	M	M	M	L	L	L	L	L	-	L	S	S
CO5	S	M	M	M	M	L	L	L	L	L	-	L	S	S
CO6	S	M	M	M	M	L	L	L	L	L	-	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	10	10	20						50																4		15			
CO2	5			20						25			5		10						25				4			15		
CO3	5	10	20						25																2	5	10			
CO4													5	5	20						25				2	5	10			
CO5													5	5	20						25				4		10			
CO6													5	10	10						25				4	10				
Total (in %)	100						100						100						100											

## Syllabus

**INTRODUCTION:** Introduction to Machine Learning – Basic concepts, Model selection – Supervised, Unsupervised, Reinforcement Learning- Example applications – Bayesian Decision Theory

**PARAMETRIC METHODS:** Introduction, Maximum Likelihood estimation, Evaluating an estimator, Bayes' estimator, Bayes estimation of regression and classification of parameters, Parametric classification, Regression, Tuning model complexity, Model selection procedures - Demonstrations

**MULTIVARIATE METHODS:** Multivariate data, Parameter estimation, Estimation of missing values, Multivariate normal distribution, Multivariate classification, Tuning complexity, Discrete features, Multivariate regression - Demonstrations

**DIMENSIONALITY REDUCTION:** Introduction, Subset Selection, Principal Components Analysis, Multidimensional Scaling, Linear Discriminant Analysis – Demonstrations

**CLUSTERING:** k- Means Clustering, Expectation Maximization

**NON-PARAMETRIC METHODS:** Introduction, Non-parametric density estimation, Generalization to multivariate data, Non-parametric classification, Non-parametric regression - Demonstrations

**LINEAR DISCRIMINATION:** Generalizing linear model, Linear discriminant, Pairwise separation, Gradient descent, Logistic discrimination, Discrimination by regression - Demonstrations

**KERNEL METHODS :** Dual Representations, Constructing Kernels, Radial Basis Function networks, Gaussian processes for Regression and classification – Laplace approximation, Maximum margin classifiers (SVM for classification and regression)

## Reference Books & web resources

1. Ethem Alaypadin, "Introduction to Machine Learning", Fourth Edition, MIT Press, 2020 (Chapters 1,2,3,4,5,6,7,8,10)
2. Christopher M. Bishop, "Pattern Recognition and Machine Learning", Springer, 2013 (Chapters 6,7)
3. Kevin P. Murphy, "Machine Learning: A probabilistic perspective", MIT Press, 2012.
4. Trevor Hastie, Robert, Jerome, "The elements of statistical learning: Datamining, Inference and Prediction", 2<sup>nd</sup> Edition, 2009.
5. Aurelien Geron, "Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow" , O'Reilly Media, 2022

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>INTRODUCTION TO MACHINE LEARNING</b>	
1.1	Basic concepts, types of learning	1

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
1.2	Model Selection- Supervised	1
1.3	Unsupervised and reinforcement learning	1
1.4	Applications of Machine Learning	1
1.5	Bayesian Decision Theory	2
<b>2</b>	<b>PARAMETRIC METHODS</b>	
2.1	Introduction, Maximum Likelihood estimation	1
2.2	Evaluating an estimator, Bayes' estimator	2
2.3	Parametric classification, Regression	2
2.4	Tuning model complexity	1
2.5	Model selection procedures	1
2.6	Demonstrations	1
<b>3</b>	<b>MULTIVARIATE METHODS</b>	
3.1	Multivariate data, Parameter estimation	1
3.2	Estimation of missing values, Multivariate normal distribution	1
3.3	Multivariate classification, Tuning complexity	1
3.4	Discrete features, Multivariate regression	2
3.5	Demonstrations	1
<b>4</b>	<b>DIMENSIONALITY REDUCTION</b>	
4.1	Introduction, Subset Selection	1
4.2	Principal Components Analysis	1
4.3	Multidimensional Scaling	1
4.4	Linear Discriminant Analysis	1
4.5	Demonstrations	1
<b>5</b>	<b>CLUSTERING</b>	
5.1	k- Means Clustering	2
5.2	Expectation Maximization	1
5.3	Demonstrations	1
<b>6</b>	<b>NON-PARAMETRIC METHODS</b>	

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
6.1	Introduction, Non-parametric density estimation	2
6.2	Generalization to multivariate data	1
6.3	Non-parametric classification, Non-parametric regression	1
6.4	Demonstrations	1
<b>7</b>	<b>LINEAR DISCRIMINATION</b>	
	Generalizing linear model, Geometry of linear discriminant	1
7.1	Pairwise separation, Gradient descent	1
7.2	Logistic discrimination	1
7.3	Discrimination by regression	1
7.4	Demonstration	1
<b>8</b>	<b>KERNEL METHODS</b>	
8.1	Dual Representations, Constructing Kernels	1
8.2	Radial Basis Function networks	1
8.3	Maximum margin classifiers (SVM for classification and regression)	2
8.4	Gaussian processes	2
8.5	Laplace approximation	1
	<b>Total</b>	<b>46</b>

**Course Designer(s):**

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**23DS550****COMPUTER NETWORKS**

Category L T P Credit

PCC 4 0 0 4

**Preamble**

The course aims to provide an understanding of computer networks architecture, various technologies available to build a network and protocols in use at different levels of network layers stack. An overview of global Internet, Internet applications and introduction to Network simulation is also provided.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Apply the fundamentals of networking, including physical transmission media and LAN/WAN technologies (Ethernet, IEEE 802.11), to establish network connectivity.	Apply
CO2	Apply data link layer protocols, including CSMA/CD, flow, and error control techniques, for reliable data transfer.	Apply
CO3	Analyze IP addressing, subnetting, and routing protocols (RIP, OSPF) to enable efficient internetworking, and evaluate network performance metrics like latency and throughput.	Analyze
CO4	Examine transport layer protocols (TCP/UDP) with emphasis on connection management, flow, and congestion control.	Analyze
CO5	Utilize application layer protocols (HTTP, FTP, DNS) for providing essential end-user network services.	Apply
CO6	Use cryptographic techniques to secure data communication and ensure network integrity.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	M	L	S	L	L	L	L	L	L	L	L	L
CO2	S	S	M	L	S	L	L	L	L	L	L	L	L	L
CO3	S	S	S	M	M	L	L	L	L	M	L	L	L	M
CO4	S	S	S	M	M	L	L	L	L	M	L	L	L	M
CO5	S	S	M	L	S	L	L	L	L	L	L	L	L	M
CO6	S	S	M	L	S	L	L	L	S	L	L	L	M	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	10	10	10						20																4	5	10			
CO2	10	10	10						30																4	5	10			
CO3		20	10	10					30	20															4	5		10		
CO4													5	10	10	10					30	20			4	5		10		
CO5													5	10	20						30				2		10			
CO6													10	10	10						20				2		10			
Total (in %)	100						100						100						100											

## Syllabus

**NETWORK ARCHITECTURE AND COMPONENTS** – ISO/OSI architecture – Functionalities of OSI Layers - TCP/IP architecture – Types of Networks (LAN, WAN, VPN, VLAN) – Network Components (NIC, Bridges, Switches, Routers, Hubs, Gateways)

**NETWORK TECHNOLOGIES TO CONNECT HOSTS** – LAN Technologies (Ethernet, Token Ring) WAN Technologies (Circuit switching- ISDN , Packet Switching – Frame Relay - (IEEE802.11) – Multiple Access – CSMA/CD – Flow control techniques – Error control techniques.

**INTERNETWORKING** – IPv4 – Global Addresses – ARP – DHCP – ICMP – Routing – Intra domain Routing algorithms (RIP, OSPF) – Subnetting – Classless Addressing – Inter domain routing – Ipv6 – Multicasting – VoIP.

**PROTOCOL STACK** – End to End protocols [TCP (segment format, connection establishment & Termination, Congestion control, Flow Control), UDP] - Application layer protocols (SMTP, IMAP, POP, HTTP, FTP, DNS, Telnet).

**NETWORK PERFORMANCE AND SIMULATION** – Bandwidth – latency – Throughput - Jitter – Delay

**NETWORK SECURITY** – Security goals, Attacks, Services and Techniques. Confidentiality-Symmetric-Key Ciphers, Asymmetric-Key Ciphers. Message Integrity, Message Authentication, Digital Signature, Entity Authentication, Key Management.

## Reference Books & web resources

1. Behrouz A. Foruzan, " Data Communication and Networking", Tata McGraw Hill, Sixth Edition, 2022. (Chapters 1 -3, 9-13, 15, 17,18, 20, 21, 22, 24-26,31)
2. William Stallings, " Data and Computer Communications", Pearson Education Ninth Edition 2013.
3. Larry L. Peterson and Bruce S. Davie, " Computer Networks – A systems Approach" Fifth Edition, Morgan Kaufmann Publishers,2011

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>NETWORK ARCHITECTURE AND COMPONENTS</b>	
1.1	ISO/OSI architecture	1
1.2	Functionalities of OSI Layers	2
1.3	TCP/IP architecture	1
1.4	Types of Networks (LAN, WAN,	2
1.5	VPN, VLAN)	2
<b>2</b>	<b>NETWORK TECHNOLOGIES TO CONNECT HOSTS</b>	
2.1	LAN Technologies (Ethernet, Token Ring)	1
2.2	WAN Technologies	1

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
2.3	Circuit switching- ISDN	2
2.4	Packet Switching – Frame Relay	2
2.5	(IEEE802.11) – Multiple Access – CSMA/CD	1
2.6	Flow control techniques	2
2.7	Error control techniques.	2
<b>3</b>	<b>INTERNETWORKING</b>	
3.1	IPv4 – Global Addresses	1
3.2	ARP – DHCP – ICMP- Routing	2
3.3	Intra domain Routing algorithms (RIP, OSPF)	2
3.4	Subnetting – Classless Addressing – Inter domain routing	2
3.5	Ipv6 – Multicasting	1
3.6	VoIP.	1
<b>4</b>	<b>PROTOCOL STACK</b>	
4.1	End to End protocols	1
4.2	TCP (segment format, connection establishment & Termination, Congestion control, Flow Control),	2
4.3	UDP .	1
4.4	Application layer protocols	1
4.5	SMTP, IMAP	2
4.6	POP, HTTP	2
4.7	FTP, DNS, Telnet	2
<b>5</b>	<b>NETWORK PERFORMANCE AND SIMULATION</b>	
5.1	Bandwidth – latency – Throughput	1
5.2	Jitter – Delay	1
<b>6</b>	<b>NETWORK SECURITY</b>	
6.1	Security goals, Attacks, Services and Techniques	1
6.2	Confidentiality- Symmetric-Key Ciphers, Asymmetric-Key Ciphers	1
6.2	Message Integrity, Message Authentication, Digital Signature, Entity Authentication, Key Management	2
	<b>Total</b>	<b>45</b>

**Course Designer:**

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**23DS570****WEB TECHNOLOGY LAB**

Category	L	T	P	Credit
PCC	0	0	4	4

**Preamble**

This lab course aims to provide hands-on experience in building and deploying dynamic web applications, enabling students to design and implement interactive websites using modern web technologies. Students will learn client-server architecture, web development frameworks, and database connectivity essential for data-driven applications in the field of Data Science.

**Prerequisite**

- 23DS280 - Object Oriented Programming Lab
- 23DS480 – Java Programming Lab

**Course Outcomes**

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

CO1	Apply modern web technologies to design and implement responsive static web pages using HTML5, CSS3, and JavaScript.	Apply
CO2	Apply client-side scripting techniques to create dynamic and interactive user interfaces with JavaScript and its frameworks.	Apply
CO3	Apply server-side programming to develop and deploy web applications using JSP and PHP	Apply
CO4	Apply asynchronous JavaScript, Servlets techniques to build real-time web applications with smooth user experiences.	Apply
CO5	Create a full-stack web application integrating front-end and back-end technologies with a database to manage user data and interactions.	Create
CO6	Create an interactive data dashboard or web application by integrating data visualization tools and machine learning models to display real-time analytics and predictions.	Create

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO2	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO3	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO4	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO5	S	S	S	S	S	S	S	S	S	S	S	S	S	S
CO6	-	-	-	-	L	L	L	M	S	L	-	L	S	S

S- Strong; M-Medium; L-Low

**List of Experiments**

- **Introduction to HTML5 and CSS3:** Designing static web pages with responsive design using CSS frameworks.
- **Client-Side Scripting with JavaScript:** Implement form validation, interactive web components using JavaScript DOM manipulation.
- **Introduction to JavaScript Libraries and Frameworks:** Using jQuery and other libraries for creating dynamic content and React requests.
- **Servlets - Handling HTTP Requests and Responses - Form Data Handling in Servlets - Session Management in Servlets**

- **JSP** - JSP scriptlets, expressions, and directives - JSP Implicit Objects - Error Handling in Servlets and JSP - Database Connectivity with JSP
- **PHP**: Setting up a local server environment (XAMPP, WAMP, or LAMP) - HTML forms and form handling in PHP – PHP with MySQL.
- Integrating PHP and JSP/Servlets with React.

#### Reference Books & web resources

1. Robin Nixon, "Learning PHP, MySQL, JavaScript, CSS & HTML5", O'Reilly Media, 2018.
2. David Flanagan, "JavaScript: The Definitive Guide", O'Reilly Media, 2020.
3. Ethan Brown, "Web Development with Node and Express", O'Reilly Media, 2019.
4. Eric Freeman, Elisabeth Robson, "Head First HTML and CSS", O'Reilly Media, 2012.
5. Brad Dayley, "Node.js, MongoDB, and Angular Web Development", Addison-Wesley, 2017.
6. [Mozilla Developer Network \(MDN\) Web Docs](#) – Comprehensive documentation for web technologies (HTML, CSS, JavaScript).
7. [W3Schools](#) – A popular platform for web development tutorials.
8. [Stack Overflow](#) – Community-driven Q&A for troubleshooting code.
9. React Documentation – Official guide to learning React.
10. Node.js Documentation – Official Node.js API documentation.

#### Course Designer:

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**23DS580****MINI PROJECT**

Category	L	T	P	Credit
EEC	0	0	6	3

**Preamble**

The primary objective of this mini project is to enhance practical skills in data manipulation, statistical analysis, and machine learning while developing a deeper understanding of relevant concepts or theories.

**Prerequisite**

- Nil

**Course Outcomes**

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

CO1	Identify and describe the real time problem statement	Analyze
CO2	Create the architecture and various design diagrams such as ER Diagram, DFD, Use case diagram, User Interface Design.	Analyze
CO3	Apply algorithms or methodologies to real-world problems in the project.	Apply
CO4	Design a prototype or model based on findings.	Create
CO5	Develop a comprehensive report or presentation that integrates all project aspects.	Create

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	-	S	-	S	S	-	M	L	M	S	M	L	M	S
CO2	S	S	S	L	S	S	L	L	S	S	L	L	S	S
CO3	M	S	S	S	S	M	M	S	S	S	S	S	S	S
CO4	M	S	S	M	S	M	M	S	S	S	S	L	S	S
CO5	M	S	M	M	M	M	M	S	S	S	M	M	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	Review-I	Review-II	Review-III	Terminal
<b>Bloom's level</b>				
CO1	10	10	10	10
CO2	10	10	10	10
CO3	20	20	20	20
CO4	20	20	20	20
CO5	20	20	20	20
CO6	20	20	20	20
<b>Total (in %)</b>	100	100	100	100

**Syllabus**

**INTRODUCTION MODULE** - Title, Abstract, Introduction, Problem Statement in local business sectors, Objectives,

**LITERATURE REVIEW MODULE** - Existing System, Proposed System, Modules Split-up and Gantt Chart, Gaps in Current system

**Design and Architecture Module** - Architecture Diagram, ER Diagram, DFD, Use case diagram (if necessary), User Interface Design

**IMPLEMENTATION MODULE** - Development Environment Setup, Coding-Key Algorithms/Functions (100% of code implementation)

**TESTING MODULE** - Testing Strategies (Unit, Integration, System), Test Cases and Results,

**RESULTS MODULE** - Data Analysis, Performance Metrics, Visualization of Results (Charts/Graphs)

**DISCUSSION MODULE** - Interpretation of Results, Comparison with Expected Outcomes, Limitations of the Study

**CONCLUSION MODULE** - Summary of Findings, Implications of the Work, Suggestions for Future Work

**DOCUMENTATION MODULE** - User Manual, Technical Documentation, Code Comments and Explanation with screenshots

**REFERENCES MODULE** - Citations

#### Reference Books & web resources

1. Gregory M.Horine, "Project Management Absolute Beginner's Guide", QUE, 5th Edition, 2022
2. GitHub: [github.com](https://github.com)
3. Kaggle: [kaggle.com](https://kaggle.com).

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**23DS610****DEEP LEARNING**

Category L T P Credit

PCC 4 0 0 4

**Preamble**

This course aims to provide a comprehensive understanding of deep learning concepts and architectures, including CNNs, RNNs, and generative models. It emphasizes key techniques in optimization, regularization, and hyper parameter tuning for effective model evaluation. Students will explore practical applications of deep learning across various domains.

**Prerequisite**

- 23DS540- Machine Learning.

**Course Outcomes**

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

CO1	Identify the foundational concepts of deep learning, including neural networks, activation functions, and loss functions.	Apply
CO2	Design convolutional neural networks (CNNs), including their architectures and applications.	Apply
CO3	Apply regularization techniques and optimization methods to improve model performance, incorporating strategies such as data augmentation for enhanced results.	Apply
CO4	Utilize hyper parameter tuning and model evaluation methods, emphasizing the role of optimization techniques in refining deep learning models for better accuracy and efficiency.	Apply
CO5	Design recurrent neural networks (RNNs) and their variants for effective sequence modeling.	Apply
CO6	Construct generative models and transformers for data generation and attention mechanisms.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	S	M	S	M	-	L	L	-	L	-	S	S
CO2	S	S	S	M	S	M	L	L	L	L	L	L	S	S
CO3	S	S	S	L	S	-	-	-	-	-	-	-	S	M
CO4	S	S	S	L	S	-	-	-	-	-	-	-	S	M
CO5	S	S	S	M	S	M	L	L	L	L	L	L	S	S
CO6	S	S	S	M	S	M	L	L	L	-	L	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	5	10	20						25																2	5	15			
CO2	5	10	25						50																2	5	15			
CO3	5	10	10						25																2		10			
CO4											5	10	15												2		10			
CO5											5	10	20												2	5	10			
CO6											5	10	20												5	10				
Total (in %)	100						100						100						100											

## Syllabus

**INTRODUCTION TO DEEP LEARNING:** Overview of Deep Learning: Motivation and Applications - Neural Networks and Biological Neurons - Deep Feedforward Networks - Backpropagation of Neural Networks - Activation Functions: ReLU, Leaky ReLU, Swish, etc. - Loss and Cost Functions.

**CONVOLUTIONAL NEURAL NETWORKS (CNNs):** Convolution and Pooling Operations - Backpropagation in CNNs - CNN Architectures: LeNet, AlexNet, VGG, ResNet, EfficientNet - Transfer Learning and Fine-Tuning Pretrained Networks.

**REGULARIZATION AND GENERALIZATION IN DEEP LEARNING:** Overfitting and Underfitting: Concepts and Impacts - Dropout - Batch Normalization and Layer Normalization - L1 and L2 Regularization Techniques - Data Augmentation Strategies: Geometric Transformations, Color Space Augmentation, Random Erasing and Noise injection.

**OPTIMIZATION TECHNIQUES FOR DEEP LEARNING:** Gradient Descent: Theory and Applications - Variants (SGD, Momentum, RMSprop, Adam, AdamW) - Advanced Optimization Techniques: Gradient Clipping, Learning Rate Scheduling, Gradient Accumulation, Early Stopping.

**HYPER PARAMETER TUNING AND EVALUATION METHODS:** Hyper parameter Tuning (Grid Search, Random Search, Bayesian Optimization) - Model Evaluation Methods: Train/Test Split, Cross-Validation - Evaluation Metrics: Accuracy, MAE, MSE, Confusion Matrix, Precision-Recall, F1-Score, ROC-AUC.

**SEQUENCE MODELING AND TEMPORAL DATA:** Recurrent Neural Networks (RNNs): Architecture and Use Cases - Backpropagation in RNNs - Long Short-Term Memory (LSTM) and Gated Recurrent Units (GRU) - Backpropagation Through Time (BPTT) for LSTMs- Introduction to Bidirectional RNNs.

**GENERATIVE MODELS AND TRANSFORMERS:** Generative Adversarial Networks (GANs) and Variants: WGAN, CycleGAN - Variational Autoencoders(VAEs) - Applications of Generative Models in Data Augmentation and Image Generation - Introduction to Attention Mechanisms and Transformers: Vision Transformers (ViT) and Efficient Vision Models.

## Reference Books

1. Ian Goodfellow, Yoshua Bengio, and Aaron Courville, "Deep Learning" (Adaptive Computation and Machine Learning Series), MIT Press, 2016.
2. Aurélien Géron, "Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow", O'Reilly Media, 2019.
3. François Chollet, "Deep Learning with Python", Manning Publications, 2021.
4. Charu Aggarwal, "Neural Networks and Deep Learning: A Textbook", Springer, 2023.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>INTRODUCTION TO DEEP LEARNING</b>	
1.1	Overview of Deep Learning: Motivation and Applications	1
1.2	Neural Networks and Biological Neurons	1
1.3	Deep Feedforward Networks	1

Module No.	Topic	No. of Periods
1.4	Backpropagation of Neural Networks	2
1.5	Activation Functions (ReLU, Leaky ReLU, Swish, etc.)	1
1.6	Loss and Cost Functions	1
<b>2</b>	<b>CONVOLUTIONAL NEURAL NETWORKS (CNNs)</b>	
2.1	Convolution and Pooling Operations	1
2.2	Backpropagation in CNNs	2
2.3	CNN Architectures: LeNet, AlexNet, VGG, ResNet, EfficientNet	2
2.4	Transfer Learning and Fine-Tuning Pretrained Networks	2
<b>3</b>	<b>REGULARIZATION AND GENERALIZATION IN DEEP LEARNING</b>	
3.1	Overfitting and Underfitting: Concepts and Impacts	1
3.2	Dropout	1
3.3	Batch Normalization and Layer Normalization	1
3.4	L1 and L2 Regularization Techniques	1
3.5	Data Augmentation Strategies- Geometric Transformations, Color Space Augmentation, Random Erasing and Noise injection.	2
<b>4</b>	<b>OPTIMIZATION TECHNIQUES FOR DEEP LEARNING</b>	
4.1	Gradient Descent: Theory and Applications	1
4.2	Variants (SGD, Momentum, RMSprop, Adam, AdamW)	2
4.3	Advanced Optimization Techniques: Gradient Clipping, Learning Rate Scheduling, Gradient Accumulation, Early Stopping.	2
<b>5</b>	<b>HYPER PARAMETER TUNING AND EVALUATION METHODS</b>	
5.1	Hyper parameter Tuning (Grid Search, Random Search, Bayesian Optimization)	2
5.2	Model Evaluation Methods: Train/Test Split, Cross-Validation.	1
5.3	Evaluation Metrics: Accuracy, MAE, MSE, Confusion Matrix, Precision-Recall, F1-Score, ROC-AUC.	2
<b>6</b>	<b>SEQUENCE MODELING AND TEMPORAL DATA</b>	
6.1	Recurrent Neural Networks (RNNs): Architecture and Use Cases	1
6.2	Backpropagation in RNNs	1
6.3	Long Short-Term Memory (LSTM) and Gated Recurrent Units (GRU)	2
6.4	Backpropagation Through Time (BPTT) for LSTMs	1
6.5	Introduction to Bidirectional RNNs.	1

Module No.	Topic	No. of Periods
<b>7</b>	<b>Generative Models and Transformers</b>	
7.1	Generative Adversarial Networks (GANs) and Variants (WGAN, CycleGAN)	2
7.2	Variational Autoencoders (VAEs)	1
7.3	Applications of Generative Models in Data Augmentation and Image Generation	2
7.4	Introduction to Attention Mechanisms and Transformers: Vision Transformers (ViT) and Efficient Vision Models.	3
	<b>Total</b>	<b>44</b>

### Course Designer

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**23DS620****DATA MINING**

Category	L	T	P	Credit
PCC	4	0	0	4

**Preamble**

This course provides a comprehensive introduction to data mining, focusing on data pre-processing, similarity measures, classification, clustering, stream mining, and sequence pattern discovery to equip students with practical data analysis and predictive modelling skills.

**Prerequisite**

- 23DS220: Applied Statistics
- 23DS340: Database Systems

**Course Outcomes**

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

CO1	Explain the architecture of data mining process, issues and importance	Understand
CO2	Describe the types of data attributes and apply basic statistical measures to analyze data distribution and variability.	Apply
CO3	Apply suitable data pre-processing methods for the given dataset.	Apply
CO4	Apply proximity measures, to assess data similarity and dissimilarity, and implement association rule mining techniques	Apply
CO5	Apply decision tree algorithms, ensemble techniques, and clustering methods to perform classification, prediction, and data segmentation in various contexts.	Apply
CO6	Apply stream mining techniques and algorithms, and implement sequence mining methods to discover patterns and analyze multivariate time series data	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	L	L	L	M	-	-	-	L	L	L	L	M	M
CO2	S	S	L	M	M	-	-	-	L	L	L	L	M	S
CO3	M	M	M	M	S	-	-	-	M	L	M	L	M	S
CO4	M	S	M	M	S	-	-	-	M	L	M	L	M	S
CO5	M	S	S	M	S	-	-	-	M	L	M	L	S	S
CO6	M	S	S	M	S	-	-	-	M	L	M	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	10	10							20																		4	6		
CO2	10	10	20						40																		4	6	10	
CO3	10	10	20						40																		6	10		
CO4													10		10											4		10		
CO5													10	10	20											4	6	10		
CO6													10	10	20											4	6	10		
Total (in %)	100						100						100						100											

## Syllabus

**INTRODUCTION:** Motivation for Data Mining - Data Mining Issues -Importance - Data Mining from a Database Perspective -Classification of Data Mining Systems – Major issues in Data Mining.**Data Pre-Processing:**Types of data - Data cleaning – Aggregation - Sampling – Feature subset selection–wrapper and filter methods.

**DATA OBJECTS AND ATTRIBUTE TYPES:**Attribute, Nominal Attributes,Binary Attributes, Ordinal Attributes,Numeric Attributes,Discrete versus Continuous Attribute, Introduction to Basic Statistical Descriptions of Data.

**Measuring Data Similarity and DISSIMILARITY:** Data Matrix versus Dissimilarity Matrix, Proximity Measures for Nominal Attributes, Binary Attributes,Ordinal AttributesDissimilarity of Numeric Data, Minkowski Distance, Dissimilarity for Attributes of Mixed Types, Cosine Similarity. **Association Rules:** Large Itemsets, Market Basket Analysis,Apriori Algorithm, Sampling Algorithm, Partitioning, Pattern Mining: Constraint-based pattern mining, Mining sequential patterns.

**DECISION TREE-BASED ALGORITHMS:** ID3, C 4.5 and C5.0, CART, **Ensemble of classifiers:** Classification – Prediction – Voting, Bagging, Boosting, Stacking, Cascading, Random forest, introduction to Semi Supervised learning.

**CLUSTERING:****Hierarchical Algorithms-**Agglomerative Algorithms, Divisive Clustering. **PartitionalAlgorithms:** Minimum Spanning Tree, K -Means Clustering, K-Medoids, Nearest Neighbor Algorithm, PAM Algorithm. **Clustering Large Databases:** BIRCH, DBSCAN, CURE Algorithm.

**MINING DATA STREAMS:** Challenges–Stream data model – Sampling data in a stream, Filtering streams: The Bloom Filter- Counting frequency items in a stream-The Count-Distinct Problem, TheFlajolet-Martin Algorithm.

**SEQUENCE MINING:** Characteristics of Sequence Data, Problem Modelling, Sequential Pattern Discovery, Timing Constraints, Multivariate Time Series (MVTs) Mining: Importance of MVTs data - Sources of MVTs data - Mining MVTs data.

**CASE STUDIES:** Web Mining, Spatial Mining, Graph Mining, Temporal Mining.

## Reference Books & Web Resources

1. Jiwei Han and Micheline Kamber- “Data Mining – Concepts and Techniques”, Morgan Kaufmann,4<sup>th</sup> edition, 2022.
2. Tan, Steinbach, Kumar,- “Introduction to Data Mining”, Pearson Education, 2014.
3. AnandRajaraman and Jeffrey Ullman – “Mining Massive Data sets”, Cambridge University Press, 2014.
4. Giovanni Seni, John Elder – “Ensemble methods in data mining: Improving accuracy through combining prediction”, Morgan &ClayPool, 2010.
5. Trevor Hastie, Robert Tibshirani, Jerome Friedman “The Elements of Statistical Learning Data Mining, Inference, and Prediction”, Second Edition
6. Ian Witten, Frank Eibe and Mark A. Hall, "Data Mining: Practical Machine Learning Tools and Techniques", Elsevier, 2011.
7. **NPTel:**[https://onlinecourses.nptel.ac.in/noc21\\_cs06/preview](https://onlinecourses.nptel.ac.in/noc21_cs06/preview)  
[https://onlinecourses.nptel.ac.in/noc19\\_cs53/preview](https://onlinecourses.nptel.ac.in/noc19_cs53/preview)

**Course Contents and Lecture Schedule**

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>1</b>	<b>INTRODUCTION</b>	
1.1	Motivation for Data Mining - Data Mining Issues -Importance - Data Mining from a Database Perspective	2
1.2	Classification of Data Mining Systems – Major issues in Data Mining.	1
1.3	<b>Data Pre-processing:</b> Types of data, Data cleaning, Aggregation, Sampling	2
1.4	Feature subset selection, wrapper and filter methods.	2
<b>2</b>	<b>DATA OBJECTS AND ATTRIBUTE TYPES</b>	
2.1	Attribute, Nominal Attributes, Binary Attributes, Ordinal Attributes, Numeric Attributes, Discrete versus Continuous Attribute.	2
2.2	<b>Measuring the Central Tendency:</b> Mean, Median, and Mode <b>Measuring the Dispersion of Data:</b> Range, Quartiles, Variance, Standard Deviation, and Interquartile Range	1
2.3	<b>Measuring Data Similarity and Dissimilarity:</b> Data Matrix versus Dissimilarity Matrix, Proximity Measures for Nominal Attributes, Binary Attributes, Ordinal Attributes	2
2.4	Dissimilarity of Numeric Data, Minkowski Distance, Dissimilarity for Attributes of Mixed Types, Cosine Similarity	2
2.5	<b>Association Rules:</b> Large Itemsets, Market Basket Analysis, Apriori Algorithm, Sampling Algorithm, Partitioning	2
2.6	<b>Pattern Mining:</b> Constraint-based pattern mining, Mining sequential patterns	2
<b>3</b>	<b>DECISION TREE-BASED ALGORITHMS</b>	
3.1	ID3, C 4.5 and C5.0, CART	2
3.2	Classification – Prediction – Voting, Bagging, Boosting	2
3.3	Boosting, Stacking, Cascading,	2
3.4	Random forest, Introduction to Semi Supervised learning	2
<b>4</b>	<b>CLUSTERING: HIERARCHICAL ALGORITHMS</b>	
4.1	Agglomerative Algorithms, Divisive Clustering	1
4.2	Minimum Spanning Tree, K -Means Clustering, K-Medoids	2
4.3	Nearest Neighbour Algorithm, PAM Algorithm.	1
4.4	<b>Clustering Large Databases:</b> BIRCH, DBSCAN, CURE Algorithm	2

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>5</b>	<b>MINING DATA STREAMS</b>	
5.1	Challenges–Stream data model –	1
5.2	Sampling data in a stream,	2
5.3	Filtering streams: The Bloom Filter	1
5.4	Counting frequency items in a stream-The Count-Distinct Problem	2
5.5	TheFlajolet-Martin Algorithm.	1
<b>6</b>	<b>SEQUENCE MINING</b>	
6.1	Characteristics of Sequence Data, Problem Modelling,	3
6.2	Sequential Pattern Discovery, Timing Constraints,	1
6.3	Multivariate Time Series (MVTs) Mining: Importance of MVTsdata -	1
6.4	Sources of MVTs data - Mining MVTs data	2
	<b>Total</b>	<b>46</b>

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23DS630

**BIG DATA SYSTEMS**

Category L T P Credit

PCC 3 1 0 4

**Preamble**

This course introduces Big Data processing techniques for business intelligence and analytics, focusing on MapReduce with Hadoop and advanced tools like Apache Spark, Pig, Hive, and HBase. Students will gain the skills to process, analyze, and derive insights from large datasets using emerging Big Data technologies.

**Prerequisite**

- 23DS340: Database Management

**Course Outcomes**

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

CO1	Appraise the business areas and real-world scenarios where BigData analytics and processing technologies can be applied.	Understand
CO2	Analyze and integrate Big Data technologies with cloud-based services, focusing on tools like Apache Spark and Google BigQueryfor large-scale data processing.	Analyze
CO3	Analyze solutions for Big Data applications using Hadoop ecosystemtools such as Pig, Hive, and HBase, while also exploring emerging technologies.	Analyze
CO4	Apply various NoSQL data models (key-value, columnar, document, and graph-based) for appropriate database management in Big Datasenarios.	Apply
CO5	Implement MapReduce and other parallel computing algorithms using Hadoop and Spark frameworks to efficiently process large datasets.	Apply
CO6	Integrate databases and perform schema mapping, ensuring information preservation and data consistency across distributed systems.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	S	S	M	S	L	S	-	S	-	S	M	M
CO2	S	S	S	M	S	-	-	-	-	-	-	-	-	S
CO3	S	S	M	S	M	M	-	-	M	-	-	-	M	S
CO4	S	M	S	M	S	-	-	-	-	-	-	-	-	S
CO5	S	S	S	S	S	-	-	-	-	-	M	-	S	S
CO6	M	S	M	M	S	-	-	-	-	-	-	-	-	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	10	20					10																		2	10				
CO2													5	10	20						25			2			15			
CO3													5	10	20							25		2			15			
CO4	5	10	20					50																	10	10				
CO5													10	20						50				2	5	10				
CO6	5	10	20						40															2	5	10				
Total (in %)	100						100						100						100											

## Syllabus

**PARALLEL AND DISTRIBUTED DATABASES:** Introduction to parallel databases - Architecture and concepts of parallel databases - Parallel query evaluation and optimization - Introduction to distributed databases (DDBMS) - DDBMS architecture and distributed database design - Distributed query processing and optimization

**INTRODUCTION TO BIG DATA:** Big Data challenges and characteristics – Types of big data (structured, semi-structured, unstructured) - NoSQL data models (Key-value, Document-based, Column-oriented, Graph-based) - SQL vs NoSQL databases – Big Data Applications (Microsoft Azure, Spark integration)

**DATA MODELING AND DATABASE INTEGRATION:** Data warehousing concepts – Schema directed data integration - Schema mapping and information preservation - Data exchange and automatic schema matching – Schema for multidimensional data models

**NOSQL DATABASES:** Key-Value Stores (Redis, Oracle Coherence) - Column-Oriented Stores (Cassandra, HBase) - Document-Oriented Stores (MongoDB, CouchDB) - Graph Databases (Neo4j, OrientDB)

**BIG DATA PROCESSING FRAMEWORKS (MAP-REDUCE):** Introduction to Apache Hadoop - Hadoop ecosystem - Concepts of HDFS - MapReduce programming model - Pig, Hive, Spark, Yarn,

**INTRODUCTION TO DATA ANALYTICS:** Overview of Data Analytics - Types of analytics (Descriptive, Diagnostic, Predictive, Prescriptive) and their applications - Basics of Statistical Analysis for Big Data - Analytics tools (Apache Spark, Google BigQuery, Kafka) - Machine Learning for Big Data (linear regression with Spark or BigQuery)

## Reference Books & web resources

1. White, T. (2012). "Hadoop: The definitive guide". O'Reilly Media, Inc. (Chapter: 1, 3, 4, 16, 17 & 19)
2. Elmasri, R., & Navathe, S. B. (2016). "Fundamentals of database systems seventh edition". (Chapter: 18, 23 & 29)
3. Acharya, S., & Chellappan, S. (2019). "Big data and analytics". Wiley. (Chapter: 1, 2, 4, 5, 6, 7, 8 & 12)
4. Warren, J., & Marz, N. (2015). "Big Data: Principles and best practices of scalable realtime data systems". Simon and Schuster.
5. EMC Education Services (Ed.). (2014). "Data science and big data analytics: discovering, analyzing, visualizing and presenting data". John Wiley & Sons.
6. Erl, T., Khattak, W., & Buhler, P. (2016). "Big data fundamentals: concepts, drivers & techniques". Prentice Hall Press.
7. Ryza, S., Laserson, U., Owen, S., & Wills, J. (2017). "Advanced analytics with spark: patterns for learning from data at scale". O'Reilly Media, Inc.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>PARALLEL AND DISTRIBUTED DATABASES</b>	
1.1	Introduction to parallel databases	1
1.2	Architecture and concepts of parallel databases	1

1.3	Parallel query evaluation and optimization	1
1.4	Introduction to distributed databases (DDBMS)	1
1.5	DDBMS architecture and distributed database design	1
1.6	Distributed query processing and optimization	1
<b>2</b>	<b>INTRODUCTION TO BIG DATA</b>	
2.1	Big Data challenges and characteristics	1
2.2	Types of big data (structured, semi-structured, unstructured)	1
2.3	NoSQL data models (Key-value, Document-based, Column-oriented, Graph-based)	2
2.4	SQL vs NoSQL databases	1
2.5	Big Data Applications (Microsoft Azure, Spark integration)	2
<b>3</b>	<b>DATA MODELING AND DATABASE INTEGRATION</b>	
3.1	Data warehousing concepts	1
3.2	Schema directed data integration	1
3.3	Schema mapping and information preservation	1
3.4	Data exchange and automatic schema matching	1
3.5	Schema for multidimensional data models	2
<b>4</b>	<b>NOSQL DATABASES</b>	
4.1	Key-Value Stores (Redis, DynamoDB)	2
4.2	Column-Oriented Stores (Cassandra, HBase)	2
4.3	Document-Oriented Stores (MongoDB, CouchDB)	2
4.4	Graph Databases (Neo4j, OrientDB)	2
<b>5</b>	<b>Big Data Processing Frameworks (Map-Reduce)</b>	
5.1	Introduction to Apache Hadoop and Hadoop ecosystem	1
5.2	Concepts of HDFS	1
5.3	MapReduce programming model	1
5.4	Pig, Hive, Spark, Yarn	5
<b>6</b>	<b>Introduction to Data Analytics</b>	
6.1	Overview of Data Analytics and Types of analytics (Descriptive, Diagnostic, Predictive, Prescriptive)	2
6.2	Basics of Statistical Analysis for Big Data	2
6.3	Analytics tools (Apache Spark, Google BigQuery, Khafka)	3
6.4	Machine Learning for Big Data (linear regression with Spark or BigQuery)	3
	<b>Total</b>	<b>45</b>



<b>23DS640</b>	<b>ETHICS FOR DATA SCIENCE</b>	Category	L	T	P	Credit
		PCC	3	0	0	3

**Preamble**

This course will enable students to recognize the ethical issues when applying data science to real world problems and learn about the ethical aspects of data science, including privacy, plagiarism, intellectual property rights, piracy, security and confidentiality.

**Prerequisite**

- NIL

**Course Outcomes**

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

CO1	Understand key ethical concepts related to data science, including the balance between right and wrong in data-driven decision-making.	Understand
CO2	Explain privacy concerns and ethical issues surrounding data collection and usage, focusing on mechanisms to protect individual rights.	Understand
CO3	Describe ethical frameworks to address privacy and fairness issues in data preprocessing, ensuring responsible use of data.	Apply
CO4	Implement ethical modeling techniques to develop fair, transparent, and accountable predictive models.	Apply
CO5	Explain ethical evaluation techniques to interpret results and ensure transparency and fairness in data-driven systems.	Apply
CO6	Use ethical principles to manage the deployment of data science solutions, considering access rights and fair treatment for different groups.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	-	-	-	-	S	-	-	M	-	-	-	M	L
CO2	M	-	-	-	-	S	-	-	M	-	-	-	S	M
CO3	M	M	S	M	M	S	M	-	M	-	-	-	S	S
CO4	M	M	S	M	M	S	M	-	M	-	-	-	S	S
CO5	M	M	M	S	M	S	M	M	M	-	-	-	S	S
CO6	M	M	S	M	M	S	M	-	M	-	-	-	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	5	10																							5	5				
CO2	5	10																							5	5				
CO3		10	40						60				5											5	5	10				
CO4			20						40				5	10	10									5	5	10				
CO5													5	10	20					50				5	10					
CO6													5	10	20					50				5	20					
Total (in %)	100						100						100						100											

## Syllabus

**INTRODUCTION TO DATA SCIENCE ETHICS** - The Rise of Data Science (Ethics), Right and Wrong, Transparency in the Design Process: Model Cards, Data Cards. Data Science Ethics Equilibrium- The FAT Flow Framework for Data Science Ethics.

**ETHICAL DATA GATHERING**- Privacy as a Human Right- Privacy Mechanisms, Case study: Backdoors and Messaging Encryption-Dating, Happiness, and Ad.

**ETHICAL DATA PREPROCESSING** - Defining and Measuring Privacy, Re-identification - Defining and Selecting Variables Case study: Pregnancy and Face Recognition- Fair Relabelling.

**ETHICAL MODELLING:** Privacy, Preserving Data Mining, Discrimination-Aware Modelling Predicting Recidivism and Redlining, Bias and Fairness in Modelling, Comprehensible Models and Explainable AI, Including Ethical Preferences, Case Study - Self Driving Cars.

**ETHICAL EVALUATION:** Ethical Measurement, Ethical Interpretation of the Results, Case Study-Ethical Reporting.

**ETHICAL DEPLOYMENT:** Access to the System, Different Treatments for Different Predictions, Societal Impact, Case Study-Censoring Search and Face Recognition

## Reference Books & web resources

1. Tales Rachel, DAVID MARTENS , “Data Science Ethics Concepts, Techniques and Cautionary”, Oxford university press, 2021. (Chapters 1 to 7)
2. Barocas, S., Hardt, M., & Narayanan, A. (2023). “Fairness and machine learning: Limitations and opportunities”. MIT press.
3. Mike Loukides, Hilary Mason and DJ Patil, “Ethics and Data Science”, O'Reilly Media; 1<sup>st</sup> edition (25 July 2018) .
4. Martens D. “Data science ethics: Concepts, techniques, and cautionary tales”. Oxford University Press; 2022 Mar 24.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>INTRODUCTION TO DATA SCIENCE ETHICS</b>	
1.1	Basic concepts	2
1.2	Challenges	1
1.3	Opportunities	1
1.4	The Rise of Data Science (Ethics), Right and Wrong	1
1.5	Data Science Ethics Equilibrium	1
1.6	Transparency in the Design Process: Model Cards, Data Cards.	1
1.7	The FAT Flow Framework for Data Science Ethics	1
<b>2</b>	<b>ETHICAL DATA GATHERING</b>	

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
2.1	Privacy as a Human Right	1
2.2	Privacy Mechanisms	1
2.3	Need to analyze Privacy	1
2.4	Factors influencing Data Privacy	1
2.5	Case study: Backdoors and Messaging Encryption-Dating, Happiness, and Ads	1
2.6	Perception / Individual Decision Making	1
2.7	Ethics in Decision Making	1
<b>3</b>	<b>ETHICAL DATA PRE-PROCESSING</b>	
3.1	Introduction	1
3.2	Defining and Measuring Privacy	1
3.3	Re-identification	1
3.4	Defining and Selecting Variables	1
3.5	Case study: Pregnancy and Face Recognition-Fair Relabelling	1
<b>4</b>	<b>Ethical Modelling: Privacy</b>	
4.1	Preserving Data Mining	2
4.2	Discrimination-Aware Modelling Predicting Recidivism and Redlining	1
4.3	Bias and Fairness in Modelling	1
4.4	Comprehensible Models	1
4.5	Explainable AI	1
4.6	Ethical Preferences, Case Study-Self-Driving Cars	1
<b>5</b>	<b>ETHICAL EVALUATION</b>	
5.1	Ethical Measurement	1
5.2	Ethical Interpretation of the Results	1
5.3	Case Study - Ethical Reporting	1
<b>6</b>	<b>ETHICAL DEPLOYMENT</b>	
6.1	Access to the System	2
6.2	Different Treatments for Different Predictions	1
6.3	Societal Impact	1

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
6.4	Case Study: Censoring Search and Face Recognition	1
	<b>Total</b>	<b>35</b>

**Course Designer(s):**

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**23DS670****DEEP LEARNING LAB**

Category	L	T	P	Credit
PCC	0	0	4	2

**Preamble**

The purpose of this course is to introduce students to deep learning concepts, including neural networks, optimization techniques, and model evaluation. Students will gain hands-on experience using frameworks like TensorFlow and PyTorch to implement CNNs, RNNs, and generative models, while exploring real-world applications and best practices in deep learning.

**Prerequisite**

- 23DS540- Machine Learning.

**Course Outcomes**

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

CO1	Design feedforward neural networks for various regression and classification tasks using Python.	Apply
CO2	Implement convolutional neural networks (CNNs) for image classification and analysis on real-world datasets.	Apply
CO3	Apply regularization techniques to improve the generalization of deep learning models.	Apply
CO4	Execute different optimization algorithms and evaluate their effectiveness in training deep learning models.	Apply
CO5	Construct recurrent neural networks (RNNs) to model and analyze sequential data.	Apply
CO6	Demonstrate generative models, attention mechanisms and transformers for advanced applications in natural language processing and image classification.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	S	M	S	M	-	L	L	-	L	-	S	S
CO2	S	S	S	M	S	M	L	L	L	L	L	L	S	S
CO3	S	S	S	L	S	-	-	-	-	-	-	-	S	M
CO4	S	S	S	L	S	-	-	-	-	-	-	-	S	M
CO5	S	S	S	M	S	M	L	L	L	L	L	L	S	S
CO6	S	S	S	M	S	M	L	L	L	-	L	L	S	S

S- Strong; M-Medium; L-Low

**List of Experiments**

1. **Neural Networks Basics:** Implement a simple feedforward neural network from scratch using NumPy -Visualize activations and layer connections.
2. **Deep Feedforward Networks and Backpropagation:** Code backpropagation in TensorFlow or PyTorch- Train and evaluate on a simple dataset.
3. **Activation Functions and Loss Functions:** Experiment with different activation functions and visualize loss curves -Understand the impact of each activation function on network convergence.
4. **Convolutional Neural Networks (CNNs):** Build a CNN from scratch, visualize feature maps, fine-tune pretrained models like ResNet, VGG, and Hugging Face

models to compare performance metrics, and implement YOLO for object detection to visualize and evaluate detection results.

5. **Regularization and Generalization Techniques:** Implement dropout and L1/L2 regularization, analyzing their effect on overfitting - Debug and analyze training vs. validation loss behavior.
6. **Optimization Techniques:** Compare optimization algorithms and implement gradient clipping and learning rate schedules - Explore debugging techniques for model convergence.
7. **Hyperparameter Tuning:** Conduct hyperparameter tuning using Grid Search and Random Search- Analyze the impact of hyperparameters on model performance.
8. **Model Evaluation and Metrics:** Implement evaluation metrics and visualize confusion matrices and ROC curves.
9. **Sequence Modeling and Temporal Data:** Build RNN models for sequence prediction or sentiment analysis- Explore advanced use cases like multivariate time series forecasting.
10. **Generative Models (GANs and VAEs):** Implement GANs for image generation and VAEs for data representation- Analyze latent space and quality of generated samples.
11. **Attention Mechanisms and Transformers:** Implement attention mechanisms and experiment with hybrid architectures- Explore Vision Transformers (ViT) for image classification tasks.

#### **Capstone Project and Model Interoperability with ONNX:**

- Build a project that integrates deep learning concepts, incorporating explainability techniques and addressing ethical concerns. Convert models between PyTorch and TensorFlow using ONNX for model interoperability.

#### **Text Books**

1. Ian Goodfellow, Yoshua Bengio, and Aaron Courville, "Deep Learning" (Adaptive Computation and Machine Learning Series), MIT Press, 2016.
2. Aurélien Géron, "Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow", O'Reilly Media, 2019.
3. François Chollet, "Deep Learning with Python", Manning Publications, 2021.
4. Charu Aggarwal, "Neural Networks and Deep Learning: A Textbook", Springer, 2023.

#### **Course Designer:**

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**23DS680 BIG DATABASE SYSTEMS LAB**

Category	L	T	P	Credit
PCC	0	0	4	2

**Preamble**

This lab course aims to give students hands-on experience with various types of big data systems and tools, including NoSQL databases, big data processing frameworks and tools. Students will explore different data storage models such as key-value, column-oriented, document, and graph databases and learn to work with data processing and machine learning tools in big data environments.

**Prerequisite**

- 23DS380: Relational Database Lab
- 23DS370: Data Structures Lab

**Course Outcomes**

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

CO1	Understand and differentiate various NoSQL databases, including key-value, column-oriented, document, and graph databases	Understand
CO2	Demonstrate data storage and retrieval operations using Oracle Coherence, HBase, MongoDB, and Neo4j.	Apply
CO3	Apply big data processing on large datasets using Hadoop MapReduce, Apache Hive, and Apache Spark	Apply
CO4	Implement real-time data streaming and ingestion using Apache Kafka, integrate other big data processing tools.	Apply
CO5	Execute machine learning tasks using Weka and interpret the results into meaningful insights.	Analyze
CO6	Design and complete a capstone project using NoSQL databases and big data tools.	Create

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	S	S	S	M	-	S	-	-	-	-	S	M
CO2	S	S	S	S	S	-	-	-	-	M	-	-	S	S
CO3	S	S	S	S	S	-	-	-	-	M	-	-	S	S
CO4	S	S	S	S	S	M	M	-	S	S	S	S	S	S
CO5	S	S	S	S	S	M	M	-	S	S	-	S	S	S
CO6	S	S	S	S	S	-	-	-	S	S	S	S	S	S

S- Strong; M-Medium; L-Low

**List of Experiments****1. Introduction to NoSQL**

Key-Value Database - Oracle Coherence

- CRUD operations in Oracle Coherence
- Implementing caching and replication strategies

Column-Oriented Database - HBase

- Performing CRUD operations in HBase
- Working with large datasets in HBase and understanding column families and regions

Document-Oriented Database - MongoDB

- CRUD operations and complex queries
- Aggregation framework in MongoDB

Graph Database - Neo4j

- Building and querying graphs in Neo4j
- Implementing relationships, traversals, and graph algorithms

## 2. Big Data Processing with Hadoop, Spark, and Hive

Hadoop - MapReduce

- Basic programs - Hadoop, HDFS, and MapReduce
- Writing MapReduce jobs for batch processing on Hadoop

Apache Hive –ETL(Extract Transform and Load)

- Data ingestion and ETL processes in Hive

Spark - Big Data Processing

- Data processing with Spark RDDs, DataFrames, and DataSets

## 3. Data Processing with Kafka

Streaming Data Transmission

- Processes to publish and retrieve messages within Kafka

Integrating Kafka with Big Data Tools

- Kafka streaming with Spark and data ingestion to HDFS or HBase

## 4. Machine Learning with Weka

Use the Weka tool for pre-processing in Machine Learning for Big Data

- Perform Classification, clustering, and association rules on large datasets using Weka

## 5. Capstone project

Design and complete a capstone project that applies a combination of NoSQL databases and big data tools to address a complex data-driven problem.

## Reference Books & Web Resources

1. Tom White, "Hadoop: The Definitive Guide", 4th Edition, O'Reilly Media, 2015.
2. Edward Capriolo, Dean Wampler, and Jason Rutherglen, "Programming Hive", 1st Edition, Publisher: O'Reilly Media, 2012.
3. Lars George, "HBase: The Definitive Guide", 1st Edition, O'Reilly Media, 2011.
4. Neha Narkhede, Gwen Shapira, and Todd Palino, "Apache Kafka: The Definitive Guide", 1st Edition, O'Reilly Media, 2017.
5. Bill Chambers and Matei Zaharia "Spark: The Definitive Guide", 1st Edition, O'Reilly Media, 2018.
6. Nishant Neeraj, "Mastering Apache Cassandra", 1st Edition, Publishing, 2013.
7. Pramod J. Sadalage and Martin Fowler, "NoSQL Distilled: A Brief Guide to the Emerging World of Polyglot Persistence", 1st Edition, Addison-Wesley, 2012.

## Course Designer:

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**23DSP10****PROJECT WORK I**

Category	L	T	P	Credit
EEC	0	0	12	12

**Preamble**

The course is designed to enhance the professional competencies of M.Sc. Data Science students by enabling them to model, analyze, and develop data-driven systems. It focuses on defining objectives, selecting suitable analytical or computational models, and evaluating outcomes through experimental or applied research approaches. This project work provides a platform for students to apply theoretical knowledge to real-world problems, thereby strengthening their analytical thinking, problem-solving, and programming skills. Moreover, it fosters teamwork, project management, and effective communication abilities essential for success in both industrial and research environments. The course also offers students an opportunity to collaborate with potential employers, research institutions, and interdisciplinary teams, preparing them for future roles as data scientists, analysts, and researchers.

**Prerequisite**

- NIL

**Course Outcomes**

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

CO1	Model real-world data science problems by defining appropriate functions, concepts, and system architectures for the chosen domain.	Analyze
CO2	Evaluate and select suitable design alternatives considering practical constraints, ethical standards, and data governance principles.	Evaluate
CO3	Apply advanced statistical, computational, and machine learning techniques using suitable software tools to implement effective solutions.	Create
CO4	Interpret and validate results obtained from experiments or data analyses to derive meaningful insights and conclusions.	Evaluate
CO5	Demonstrate teamwork, leadership, and effective communication skills while working in industrial, societal, or research-based environments.	Apply
CO6	Prepare, document, and present technical project reports and research findings in alignment with national and international academic standards.	Create

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	L	M	M	S	M	–	–	–	–	–	–	S	L	M
CO2	L	M	L	M	M	–	–	–	–	–	–	S	L	M
CO3	L	M	L	M	L	–	–	–	M	M	–	S	L	L
CO4	L	M	L	M	L	–	–	–	M	M	–	S	L	L
CO5	L	M	L	M	L	–	–	–	M	M	–	S	L	L
CO6	L	M	L	L	L	–	M	–	L	L	L	L	L	L

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	Review 1						Review 2						Review 3						Terminal (Viva Voce)					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																								
CO1				5																		5		
CO2					5																		5	
CO3											10													10
CO4												10												10
CO5															15						5			
CO6																	15							5
Total (in %)	10						20						30						40					

**Project Requirements - First Review**

**Expectations from Students (in the Presentation)**

<b>Review-I</b>	Title
	About the organization
	Abstract
	Introduction
	Proposed System
	Modules Split-up and Gantt Chart
	Architectural Design for Proposed System
	ER Diagram, DFD, Use case diagram ( if necessary)
	Algorithms / Techniques used
	Expected outcomes
	References
	30% of code implementation

**Project Requirements - Second Review**

**Expectations from Students (in the Presentation)**

<b>Review-II</b>	Title
	Abstract
	Functionalities with description
	Risk Analysis
	Data Base Design – Table design and relationships
	Detailed Design –UML Diagrams
	Results obtained (intermediate)- Screenshots
	Unit Testing screen shots for modules completed so far
	Integration Testing screen shots
	References
	70% of code implementation

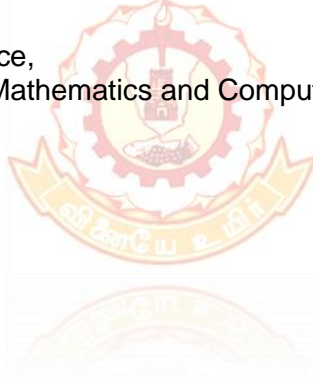
**Project Requirements – Third Review**

***Expectations from Students (in the Presentation)***

<b>Review-III</b>	Title
	Abstract
	Overall Design
	Implementation of all Modules-Screenshots
	Experimental Results
	Performance Evaluation
	Comparison with Existing system
	Unit Testing screen shots for all modules
	Integration Testing screen shots
	References
	100% of code implementation

**Course Designer:**

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**23DS810 REINFORCEMENT LEARNING**

Category	L	T	P	Credit
PCC	4	0	0	4

**Preamble**

This course introduces the core principles of Reinforcement Learning, covering topics such as Markov Decision Processes, value functions, and learning algorithms. It emphasizes both conceptual understanding and mathematical foundations, with hands-on implementation of decision-making systems through programming assignments and projects.

**Prerequisite**

- 23DS210: Theory of Probability
- 23DS220: Applied Statistics

**Course Outcomes**

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

CO1	Explain the fundamental elements of Reinforcement Learning, including evaluative feedback, rewards, returns, and the formulation of Markov Decision Processes (MDPs) and value functions.	Understand
CO2	Apply Dynamic Programming methods such as Policy Evaluation, Policy Improvement, and Value Iteration to derive optimal policies efficiently.	Apply
CO3	Implement Monte Carlo approaches for policy evaluation and control using both on-policy and off-policy learning strategies.	Apply
CO4	Utilize Temporal-Difference Learning techniques, including TD-prediction, SARSA, Q-learning, R-learning, and Actor–Critic models, for improving decision-making performance.	Apply
CO5	Employ function approximation methods such as Gradient Descent, Linear Models, and Neural Network–based techniques for value prediction and control.	Apply
CO6	Apply Deep Reinforcement Learning algorithms to develop effective solutions for complex control and decision-making problems.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	L	M	M	S	M	–	–	–	–	–	–	S	L	M
CO2	L	M	L	M	M	–	–	–	–	–	–	S	L	M
CO3	L	M	L	M	L	–	–	–	M	M	–	S	L	L
CO4	L	M	L	M	L	–	–	–	M	M	–	S	L	L
CO5	L	M	L	M	L	–	–	–	M	M	–	S	L	L
CO6	L	M	L	L	L	–	M	–	L	L	L	L	L	L

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	10	10																							4	5				
CO2	10	10	20						50																4	5	10			
CO3	10	10	20						50																4	5	10			
CO4													10	10	10						30			4	5	10				
CO5													5	10	20						30			2	5	10				
CO6													5	10	20						40			2	5	10				
Total (in %)	100						100						100						100						100					

**Syllabus**

**Reinforcement Problem:** Introduction - Elements of RL, History of RL- Evaluative feedback -Goals and rewards – Returns – Markovian Decision Problem (MDP) – Value Functions- Optimality Criterion in MDPs.

**Dynamic Programming:** Policy Evaluation- Policy Improvement- Value Iteration, asynchronous DP- Efficiency of DP.

**Monte Carlo Methods:** Policy Evaluation- Policy Improvement- On-policy and off policy Monte Carlo Controls-Incremental implementation.

**Temporal Difference Learning:** TD-prediction- Optimality of TD - Sarsa- Q-Learning – R-Learning-Actor Critic Model- Unifying Monte Carlo and TD-Traces- Games.

**Function Approximation:** Value prediction and control – Gradient Descent Methods-Linear methods – Artificial Neural Network based approximation.

**Deep RL:** PPO-DDPG- Double Q-Learning-TRPO-A3C.

**Case Studies:** RL for Game Strategy Optimization-Autonomous Robot Navigation with RL- Intelligent Traffic Signal Control-RL-Based Stock Portfolio Management.

**Reference Books & web resources**

1. Richard S. Sutton and Andrew G. Barto, *Reinforcement Learning: An Introduction*, 2nd Edition, MIT Press, 2018. (Chapters 1 to 9).
2. Maxim Lapan, *Deep Reinforcement Learning Hands-On: Apply Modern RL Methods*, 3rd Edition, Packt Publishing, 2022.
3. Miguel Morales, *Grokking Deep Reinforcement Learning*, Manning Publications, 2022.
4. Aske Plaat, *Deep Reinforcement Learning*, Springer, 2022.
5. Julien Vitay, *Deep Reinforcement Learning*, 2020.

**Course Contents and Lecture Schedule**

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>1</b>	<b>Introduction to Reinforcement Learning</b>	
1.1	Elements of RL	2
1.2	History of RL	1
1.3	Evaluative Feedback	1
1.4	Goals and Rewards, Returns	1
1.5	Markovian Decision Problem (MDP)	2
1.6	Value Functions, Optimality Criterion in MDP's	2
<b>2</b>	<b>Dynamic Programming (DP)</b>	
2.1	Policy Evaluation	2
2.2	Policy Improvement	1
2.3	Value Iteration	1
2.4	Asynchronous DP	1
2.5	Efficiency of DP	1
<b>3</b>	<b>Monte Carlo Methods</b>	
3.1	Policy Evaluation	2
3.2	Policy Improvement	1
3.3	On-policy and off policy Monte Carlo controls	2
3.4	Incremental implementation.	1
<b>4</b>	<b>Temporal Difference Learning</b>	
4.1	TD Prediction	2
4.2	Optimality of TD	1
4.3	Sarsa	1
4.4	Q Learning	1
4.5	R Learning	1
4.6	Actor, Critic Model	1
4.7	Unifying Monte Carlo and TD-Traces	1
4.8	Games	1

<b>5</b>	<b>Function Approximation</b>	
5.1	Value prediction and control	2
5.2	Gradient Descent methods	2
5.3	Linear Methods	1
5.4	Artificial Neural Network based Approximation	2
<b>6</b>	<b>Deep RL</b>	
6.1	PPO	2
6.2	DDPG	2
6.3	Double Q-Learning	1
6.4	TRPO	1
6.5	A3C	1
<b>7</b>	<b>Case Studies</b>	
7.1	RL for Game Strategy Optimization	1
7.2	Autonomous Robot Navigation with RL	1
7.3	Intelligent Traffic Signal Control	1
7.4	RL-Based Stock Portfolio Management	1
	Total	48

**Course Designer:**

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**23DS820****DATA VISUALIZATION**

Category	L	T	P	Credit
PCC	4	0	0	4

**Preamble**

This course introduces theoretical foundations and design principles of data visualization, tools for exploratory data analysis (EDA), and modern frameworks for static and dynamic visualization. It enables students to represent complex datasets, including textual, spatial, and network data, in meaningful and interactive ways to support data-driven decision-making.

**Prerequisite**

- 23DS270: Python Programming and Applied Statistics Lab
- 23DS420: Predictive Analytics
- 23DS630: Big Data Systems

**Course Outcomes**

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

CO1	<b>Outline</b> the fundamental principles and theoretical foundations of information visualization.	Understand
CO2	Apply exploratory data analysis techniques to identify trends, outliers, and relationships in datasets.	Apply
CO3	Use visualization tools to create static and interactive data visualizations from various data formats.	Apply
CO4	Develop dynamic and web-based visualizations using tools or frameworks like PowerBi/Tableau/Matplotlib.	Apply
CO5	Construct specialized visualizations such as choropleth maps and network trees to represent spatial and relational data.	Apply
CO6	Design visualization solutions for large-scale and unstructured data including text and biological sequences.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1.	L	M	M	L	S	L	L	L	-	L	-	L	M	L
CO2.	L	M	S	M	S	M	L	L	-	L	-	L	M	L
CO3	L	M	M	M	S	M	L	L	-	L	-	L	M	L
CO4	M	S	S	M	S	M	L	L	-	L	-	L	M	L
CO5	M	S	S	M	S	M	L	L	-	L	-	L	M	L
CO6	S	S	S	S	S	M	L	L	-	L	-	L	M	L

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	15	15	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	2	10	-	-	-	-
CO2	-	15	15	-	-	-	-	-	-	50	-	-	-	-	-	-	-	-	-	-	-	-	-	-	2	5	10	-	-	-
CO3	-	20	20	-	-	-	-	-	-	50	-	-	-	-	-	-	-	-	-	-	-	-	-	-	2	5	10	-	-	-
CO4	-	-	-	-	-	-	-	-	-	-	-	10	10	10	10	-	-	-	-	-	30	-	-	-	2	5	10	-	-	-
CO5	-	-	-	-	-	-	-	-	-	-	-	05	15	15	-	-	-	-	-	30	-	-	-	2	5	10	-	-	-	
CO6	-	-	-	-	-	-	-	-	-	-	-	05	-	-	30	-	-	-	-	-	40	-	-	-	-	-	20	-	-	
Total (in %)	100						100						100						100											

\* Terminal examination should cover all Course Outcomes in the appropriate Bloom's Taxonomy level.

## Syllabus

**INTRODUCTION:** Information visualization, Theoretical foundations, Information visualization types, Design principles - A framework for producing data visualization.

**EXPLORATORY DATA ANALYSIS:** EDA fundamentals, Significance of EDA, Comparing EDA with classical and Bayesian analysis, Software tools for EDA, Visual Aids for EDA Data transformation techniques-merging database, reshaping and pivoting, Transformation techniques - Grouping Datasets, Data Aggregation – Pivot tables and cross-tabulations.

**STATIC DATA VISUALIZATION:** Tools, Working with various data formats

**DYNAMIC DATA DISPLAYS:** Introduction to web based visual displays, deep visualization, collecting sensor data, visualization D3 framework - Introduction to Power BI/Tableau

**MAPS** – Introduction to building choropleth maps, Area Map and Flow maps.

**TREES** – Network visualizations – Displaying behaviour through network graphs.

**BIG DATA VISUALIZATION** – Visualizations to present and explore big data – visualization of text data and Protein Sequences

## Reference Books & web resources

1. Ware, Colin. “*Visual Thinking for Design.*” Morgan Kaufmann, 2008. (Introduction)
2. Mukhiya, Suresh Kumar and Usman Ahmed. “*Hands-On Exploratory Data Analysis with Python.*” Packt Publishing, 2020. (Exploratory Data Analysis)
3. Fry, Ben. “*Visualizing Data.*” O’Reilly Media, 2007 (Static Data Visualization)
4. Murray, Scott. “*Interactive Data Visualization for the Web.*” O’Reilly Media, 2017 (D3.js)
5. Kieran Healy. “*Data Visualization: A Practical Introduction.*” Princeton University Press, 2018. (Maps and Tree visualization)

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>INTRODUCTION</b>	
1.1	Information visualization	1
1.2	Theoretical foundations	1
1.3	Information visualization types	1
1.4	Design principles	1
1.5	A framework for producing data visualization	1
<b>2</b>	<b>EXPLORATORY DATA ANALYSIS</b>	
2.1	EDA fundamentals	1
2.2	Significance of EDA	1
2.3	Comparing EDA with classical and Bayesian analysis	1

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
2.4	Software tools for EDA	2
2.5	Visual Aids for EDA Data transformation techniques	1
2.6	Merging database	1
2.7	Reshaping and pivoting	2
2.8	Transformation techniques	1
2.9	Grouping Datasets	1
2.10	Data Aggregation	1
2.11	Pivot tables and cross-tabulations	1
<b>3</b>	<b>STATIC DATA VISUALIZATION</b>	
3.1	Tools for static data visualization	2
3.2	Working with various data formats	2
<b>4</b>	<b>DYNAMIC DATA DISPLAYS</b>	
4.1	Introduction to web based visual displays	2
4.2	Deep visualization	2
4.3	Collecting sensor data	2
4.4	Visualization D3 framework	2
4.5	Introduction to Power BI/Tableau	2
<b>5</b>	<b>MAPS</b>	
5.1	Introduction to building choropleth maps	2
5.2	Area Map	1
5.3	Flow maps	1
<b>6</b>	<b>TREES</b>	
6.1	Network visualizations	2
6.2	Displaying behaviour through network graphs.	2
<b>7</b>	<b>BIG DATA VISUALIZATION</b>	
7.1	Visualizations to present and explore big data	2
7.2	Visualization of text data and	2
7.3	Visualization Protein Sequences	2
	Total	46

**Course Designer(s):**

1. Dr.R.SIVANESAN

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Dept. of Applied Mathematics and Computational Science

**23DS830****BUSINESS ANALYTICS**

Category	L	T	P	Credit
PCC	3	1	0	4

**Preamble**

The course is designed to gain an understanding of how managers use business analytics to formulate and solve business problems and to support managerial decision making. The course familiarizes the students with the processes needed to develop and analyze business data.

**Prerequisite**

- 23DS220 – Applied Statistics
- 23DS420 – Predictive Analytics
- 23DS520 – Optimization Techniques

**Course Outcomes**

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

CO1	Apply descriptive and exploratory techniques to analyze business datasets, handle missing or outlier data, and identify relationships among variables for decision-making.	Apply
CO2	Apply statistical inference techniques, including sampling and hypothesis testing, to draw valid conclusions about business parameters.	Apply
CO3	Develop regression models to study and interpret relationships between business variables and predict outcomes	Apply
CO4	Construct forecasting models for time series business data using regression, moving averages, exponential smoothing, and seasonal models with appropriate validation	Apply
CO5	Solve optimization problems in business scenarios such as product mix, scheduling, logistics, blending, and financial planning using suitable modeling approaches	Apply
CO6	Analyze business scenarios using analytics models to make informed decisions and recommend actionable solutions.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	M	M	M	L	L	L	L	L	-	L	S	S
CO2	S	S	M	S	M	L	L	L	L	L	-	L	S	S
CO3	S	S	M	M	M	L	L	L	L	L	-	L	S	S
CO4	S	S	S	S	M	L	L	L	L	L	-	L	S	S
CO5	S	S	S	S	M	L	L	L	L	L	-	L	S	S
CO6	S	S	M	M	M	L	L	L	L	L	-	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1				Assignment 1				CAT2				Assignment2				Terminal					
	1	2	3	4	1	2	3	4	1	2	3	4	1	2	3	4	1	2	3	4	5	6
Bloom's level																						
CO1	-	10	20		-	-	25	-	-	-	-	-	-	-	-	-	2	5	10	-	-	-
CO2	5	10	20	-	-	-	50	-	-	-	-	-	-	-	-	-	2	5	15	-	-	-
CO3	5	10	20	-	-	-	25	-	-	-	-	-	-	-	-	-	2	5	10	-	-	-
CO4	-	-	-	-	-	-	-	-	5	10	25	-	-	-	30	-	2	5	10	-	-	-
CO5	-	-	-	-	-	-	-	-	5	10	25	-	-	-	50	-	2	-	15	-	-	-
CO6	-	-	-	-	-	-	-	-	-	-	-	20	-	-	-	20	-	-	-	10	-	-
Total (in %)	100				100				100				100				100					

**Syllabus**

**INTRODUCTION TO BUSINESS ANALYTICS:** Introduction to Business Analytics, Modeling and Models

**EXPLORING BUSINESS DATA:** Basic concepts, Descriptive measures for categorical and numerical values, Time series data, Outliers and Missing values, Excel tables for filtering, sorting and summarizing, Finding relationship among Variables, Elements of decision analysis, One stage and multistage decision problems

**STATISTICAL INFERENCE:** Sampling Distribution, Sample size selection, Hypothesis testing: Concepts, Hypothesis tests for a population mean and other parameters

**REGRESSION ANALYSIS IN BUSINESS CONTEXT:** Graphing Relationships, Simple Linear Regression, Multiple Regression, Modeling possibilities, Validation of the fit, Multi collinearity, Stepwise Regression

**TIME SERIES ANALYSIS AND FORECASTING:** Regression based Trend models, Random walk model, Moving Averages Forecasts, Exponential Smoothing forecasts, Seasonal models

**OPTIMIZATION MODELING:** Two-Variable Product Mix model, Sensitivity Analysis, Employee scheduling models, Blending Models, Logistics Models, Aggregate Planning models, Financial models, Integer Optimization Models, Nonlinear Optimization models

**Reference Books & web resources**

1. S. Christian Albright and Wayne L. Winston, *Analytics: Data Analysis and Decision Making*, Sixth Edition, 2021 (Chapters 1,2,3,6,7,8,9, 10, 12, 13, 14)
2. U. Dinesh Kumar, *Business Analytics: The Science of Data-Driven Decision Making*, Wiley, 2017
3. James Evan, *Business Analytics- Methods, Models, and Decisions* (2<sup>nd</sup> Edition), Pearson, 2013
4. Gert H. N. Laursen, *Business Analytics for Managers: Taking Business Intelligence Beyond Reporting*, Wiley (2<sup>nd</sup> Edition),2016

**Course Contents and Lecture Schedule**

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>1</b>	<b>Introduction</b>	
1.1	Introduction to Business Analytics, Modeling and Models	2
<b>2</b>	<b>Exploring Business Data</b>	
2.1	Basic concepts, Descriptive measures for categorical and numerical values	2
2.2	Time series data, Outliers and Missing values	2
2.3	Excel tables for filtering, sorting and summarizing	2
2.4	Finding relationship among Variables	2
2.5	Elements of decision analysis, One stage and multistage decision problems	2
<b>3</b>	<b>Statistical Inference</b>	
3.1	Sampling Distribution	1
3.2	Sample size selection	2
3.3	Hypothesis testing: Concepts	2
3.4	Hypothesis tests for a population mean and other parameters	2
<b>4</b>	<b>Regression Analysis In Business Context</b>	
4.1	Graphing Relationships	1
4.2	Simple Linear Regression	2
4.3	Multiple Regression	2
4.4	Modeling possibilities	2
4.5	Validation of the fit, Multi collinearity	1
	Stepwise Regression	2
<b>5</b>	<b>Time Series Analysis And Forecasting</b>	
5.1	Regression based Trend models, Random walk model	2
5.2	Moving Averages Forecasts	2
5.3	Exponential Smoothing forecasts, Seasonal models	2
<b>6</b>	<b>Optimization Modeling</b>	
6.1	Two-Variable Product Mix model	1
6.2	Sensitivity Analysis	2
6.3	Employee scheduling models, Blending Models	2
6.4	Logistics Models	2
6.5	Aggregate Planning models	1
6.6	Financial models	1
6.7	Integer Optimization Models	2
6.8	Nonlinear Optimization models	2
	<b>Total</b>	<b>48</b>

**Course Designer(s):**

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**23DS870****MATHEMATICAL COMPUTING  
LAB**

Category	L	T	P	Credit
PCC	0	0	2	2

**Preamble**

The Mathematical Computing Lab enables students to translate concepts from calculus, linear algebra, optimization, and numerical methods into computational solutions. This course enhances analytical thinking, mathematical modelling, and algorithmic implementation skills necessary for advanced data science practice.

**Prerequisite**

- 23DS140 PROBLEM SOLVING USING C PROGRAMMING
- 23DS520 OPTIMIZATION TECHNIQUES

**Course Outcomes**

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

CO1	Implement numerical differentiation and integration techniques to solve rate-of-change and area-estimation problems.	Apply
CO2	Apply root-finding algorithms to solve nonlinear equations in mathematical and data-oriented models.	Apply
CO3	Apply linear programming techniques to optimize problems and solve inequalities arising in engineering applications.	Apply
CO4	Apply various methods under the transportation model to determine the minimum cost of transportation.	Apply
CO5	Apply the concepts of PERT and CPM for effective decision-making and optimal project management.	Apply
CO6	Apply eigenvalue and eigenvector computation techniques for analysing multivariate data problems.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO2	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO3	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO4	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO5	S	S	S	S	S	S	S	S	S	S	S	S	S	S
CO6	-	-	-	-	L	L	L	M	S	L	-	L	S	S

S- Strong; M-Medium; L-Low

**List of Experiments**

Given below is the list of mathematical concepts using which the programs have to be developed by the students using C / C++ / JAVA / Python to address the given problems.

1. Implement numerical differentiation using forward, backward, and central difference formulas.
2. Implement numerical integration using Trapezoidal and Simpson's rules on real data.
3. Implement the Bisection Method to find roots of equations.

**4. Optimization Techniques using Linear Programming**

Write a program to solve linear programming problems using:

- Simplex Method
- Two-Phase Method
- Dual Simplex Method

**5. Transportation Model Optimization**

(a) Write a program to find the initial basic feasible solution using:

- North-West Corner Rule
- Vogel's Approximation Method

(b) Perform optimality test using the **MODI Method** to minimize transportation cost.

**6. Assignment Problem**

- Write a program to solve the assignment problem using the **Hungarian Algorithm** for optimal resource allocation.

**7. Project Scheduling and Network Analysis**

- Write a program to construct PERT and CPM networks.
- Identify the **critical path**, calculate project duration, and analyze slack times for each activity.

**8. Perform PCA on a dataset using eigen decomposition.**

**Reference Books**

1. Narsingh Deo, *Graph Theory with Applications to Engineering and Computer Science*, Prentice Hall of India, 2017.
2. Kanti Swarup, P.K. Gupta, and Man Mohan, *Operations Research*, Sultan Chand & Sons, 2020.
3. S. S. Rao, *Optimization: Theory and Applications*, New Age International Publishers, 2019.
4. Brian Kernighan and Dennis Ritchie, *The C Programming Language*, Prentice Hall, 2015.
5. Ken Black, *Business Statistics: For Contemporary Decision Making*, Wiley, 2021.
6. Richard Bronson and Govindasami Naadimuthu, *Operations Research*, McGraw Hill, 2017.
7. Hamdy A. Taha, *Operations Research: An Introduction*, Pearson Education, 2020.

**Course Designer (s):**

- |   |                |
|---|----------------|
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**23DS880****BUSINESS ANALYTICS AND  
DATA VISUALIZATION LAB**

Category	L	T	P	Credit
PCC	0	0	2	2

**Preamble**

This course enables students to translate business data into actionable insights through analytical modeling and visual storytelling. The students will learn to explore, analyze, and visualize business data using platforms such as Excel/ Tableau/Python/ Power BI, while developing an understanding of how analytics drives strategic business decisions. By the end of this course, learners will be equipped to model business scenarios, forecast outcomes, optimize decisions, and communicate findings through interactive dashboards and reports.

**Prerequisite**

- 23DS270 - Python Programming and Applied Statistics Lab
- 23DS470 - Predictive Analytics Lab

**Course Outcomes**

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

CO1	Import, clean, and summarize business datasets using suitable tools (Excel, Tableau, Power BI, Python) and compute descriptive statistics to understand data characteristics and detect outliers.	Apply
CO2	Visualize correlations, scatter plots, and pivot tables to identify key performance indicators (KPIs) and construct basic analytical dashboards	Analyze
CO3	Formulate and test business hypotheses using appropriate statistical methods and A/B testing, and interpret results effectively through dashboards.	Evaluate
CO4	Develop and compare predictive models, including multiple and logistic regression, for understanding business relationships and classifying outcomes with model performance visualization.	Evaluate
CO5	Perform time series forecasting using moving averages and exponential smoothing, visualize seasonal patterns, and interpret trends and business cycles.	Analyze
CO6	Design and integrate business optimization and decision-support dashboards by combining statistical, predictive, and forecasting insights for real-world business case studies and reporting.	Create

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO2	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO3	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO4	S	M	M	M	S	L	L	L	M	L	-	M	S	S
CO5	S	S	S	S	S	S	S	S	S	S	S	S	S	S
CO6	-	-	-	-	L	L	L	M	S	L	-	L	S	S

S- Strong; M-Medium; L-Low

## List of Experiments

Experiments shall be done on data available on Universal repositories / self-collected survey data in Excel/Tableau/Power BI/Python

1. Import, clean, and summarize business datasets using Excel; compute descriptive statistics and explore categorical and numerical data with outlier detection
2. Use appropriate platform to visualize correlations, scatter plots, and pivot tables; Create KPI dashboard
3. Formulate business hypotheses and perform statistical tests & A/B testing on business datasets and visualize in dashboards
4. Construct and compare multiple regression models, including stepwise selection approaches
5. Perform a business classification problem using logistic regression (E.g. Churn/Non churn). Interpret coefficients and visualize ROC curves
6. Apply moving averages and exponential smoothing to business time series for forecasting and visualizing in dashboards; Visualize seasonality patterns and interpret business cycles.
7. Formulate and implement application-oriented optimization models for real-world scenarios such as product mix planning and employee scheduling using appropriate linear programming methods and computational tools.
8. Combine insights from multiple analyses (statistics, regression, forecasting). Build interactive dashboards in integrating KPIs and visualizations by using filters, slicers, and dynamic parameters for storytelling.
9. End-to-End Business Case Study
  - a) Perform complete analysis on a business dataset (e.g., retail, banking, or HR analytics).
  - b) Include data cleaning, EDA, hypothesis testing, modeling, forecasting, and visualization.
  - c) Present findings as a professional analytics report or executive dashboard.

## Reference Books & web resources

1. Albright, S. C., & Winston, W. L. (2016). *Business analytics: Data analysis and decision making* (6th ed.). Cengage Learning.
2. Provost, F., & Fawcett, T. (2013). *Data science for business: What you need to know about data mining and data-analytic thinking*. O'Reilly Media.
3. Keating, B., Wilson, J. H., & John Galt Solutions. (2020). *Forecasting and predictive analytics* (7th ed.). McGraw-Hill Education
4. Wexler, S., Shaffer, J., & Cotgreave, A. (2017). *The big book of dashboards: Visualizing your data using real-world business scenarios*. Wiley.

## Course Designer(s):

1. Dr. ANITHA D, anithad@tce.edu  
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2. Dr. R. SIVANESAN rsnds@tce.edu  
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**23DS910****WEB ANALYTICS**

Category	L	T	P	Credit
PCC	3	1	0	4

**Preamble**

This course introduces fundamental and modern concepts of web analytics, focusing on data collection, user behavior analysis, and performance measurement using contemporary tools and techniques.

**Prerequisite**

- 23DS420: Database Management
- 23DS450: Predictive Analytics
- 23DS520: Web Technology

**Course Outcomes**

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

CO1	Explain the fundamental concepts of web analytics, including behaviorism, trace data, and unobtrusive data collection methods.	Understand
CO2	Apply web data collection techniques such as web server log files, page tagging, cookies, and data accuracy measures in web analytics applications. .	Apply
CO3	Apply web analytics metrics such as visitor type, visit length, navigation paths, and keyword analysis to interpret website performance.	Apply
CO4	Apply web analytics strategies including KPI identification, measuring acquisition, conversion, retention, and clickstream metrics.	Apply
CO5	Apply web analytics tools and platforms to perform content organization, visitor segmentation, campaign analysis, and commerce measurement.	Apply
CO6	Analyze web analytics data using Google Analytics features to interpret user behavior and evaluate website performance.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	-	-	-	-	-	-	-	L	-	L	M	-
CO2	S	M	-	-	M	-	-	-	-	L	-	-	M	-
CO3	M	S	M	-	M	-	-	-	-	-	-	-	M	M
CO4	M	S	M	-	L	-	-	-	-	-	-	-	M	M
CO5	-	M	M	-	M	-	-	-	L	-	-	-	M	S
CO6	-	S	M	M	M	-	-	-	-	-	-	-	M	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	5	10																							3	5				
CO2	5	10	20						50																3		12			
CO3	5	10	25						50																3	5	12			
CO4		10											5	10	15						30			3	5	12				
CO5														15	20						30			3	5	12				
CO6														10		25					20	20			5		12			
Total (in %)	100						100						100						100						100					

**Syllabus**

**The Foundations of Web Analytics: Theory and Methods.** – Introduction- Behaviorism- Behaviors-Trace Data, Unobtrusive Methods.

**Data Collection for Web Analytics-** Web server Log Files- Page Tagging, cookies in Web analytics, Understanding web analytics data accuracy.

**Web Analytics Fundamentals** – Visitor Type, Visit Length, Internal Search, Visitor Path, Top Pages, Referrers and Keyword Analysis, Errors.

**Web Analytics Strategy-** Key Performance Indicators, Web Analytics Process, Site visits – Surveys – Measuring reach – Measuring acquisition – Measuring conversion – Measuring retention – Security and privacy implications of Web analytics. **Click Stream Metrics:** Eight critical Web Metrics

**Web Analytics Tools:** Content organization tools – Process measurement tools – Visitor segmentation tools – Campaign analysis tools – Commerce measurement tools.

**Google Analytics:** Key features and capabilities – Quantitative and qualitative data - Working of Google analytics – Privacy - Tracking visitor clicks.

**Case Studies:** GA4, User Behavior Analysis, Retention Analytics, E-commerce Analytics Case Study

**Text Book**

1. Bernard J. Jansen, *Understanding User-Web Interactions via Web Analytics*, 1st Edition, Springer Cham, 2022. (Chapters 2, 4,5,6)
2. Avinash Kaushik, "Web Analytics2.0", John Wiley and Sons, 2010.(Chapter 3)
3. Web Analytics: An Hour a Day – Avinash Kaushik,2007 (Chapter 6, 7, 8, 9, 11, and 14)
4. Brian Clifton,"Advanced web metrics with Google analytics", John Wiley and Sons,2012. (Chapter 3)

**Reference Books & web resources**

1. Justin Cutroni, Google Analytics, O'Reilly, 2015.
2. Jerri L. Ledford, Joe Teixeira and Mary E. Tyler, —Google AnalyticsII, John Wiley and Sons, 2013.

**Course Contents and Lecture Schedule**

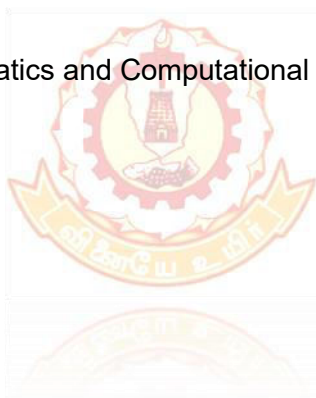
<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>1</b>	<b>Introduction</b>	
1.1	<b>Introduction Fundamentals of Web Analytics</b>	1
1.2	Behaviorism- Behaviors	1
1.3	Trace Data, Unobtrusive Methods	2
<b>2</b>	<b>Data Collection for Web Analytics</b>	
2.1	Web server Log Files	2
2.2	Page Tagging	2
2.3	cookies in Web analytics	1
2.4	Understanding web analytics data accuracy	2
<b>3</b>	<b>Web Analytics Fundamentals</b>	
3.1	Visitor Type, Visit Length	2
3.2	Internal Search Visitor Path, Top Pages	2
3.3	Referrers and Keyword Analysis	2
3.4	Errors	1
<b>4</b>	<b>Web Analytics Strategy</b>	
4.1	Key Performance Indicators	1
4.2	Web Analytics Process	2
4.3	Site visits – Surveys	1
4.4	Measuring reach – Measuring acquisition	2
4.5	Measuring conversion – Measuring retention	1
4.6	Security and privacy implications of Web analytics	1
4.7	Eight critical Web Metrics	3
<b>5</b>	<b>Web Analytics Tools</b>	
5.1	Content organization tools	2
5.2	Process measurement tools	1
5.3	Visitor segmentation tools	2
5.4	Discourse connectives	1
5.5	Campaign analysis tools	2

Module No.	Topic	No. of Periods
5.6	Commerce measurement tools	2
<b>6</b>	<b>Google Analytics:</b>	
6.1	Key features and capabilities	2
6.2	Quantitative and qualitative data	2
6.3	Working of Google analytics	2
6.4	Privacy	2
6.5	Tracking visitor clicks	1
	Total	48

**Course Designer(s):**

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**23DS920****NATURAL LANGUAGE  
PROCESSING**Category L T P Credit  
PC 3 1 0 4**Preamble**

This course aims to provide a strong theoretical foundation in Natural Language Processing, covering linguistic levels of language analysis, classical statistical methods, and modern Transformer-based architectures, enabling students to understand and critically evaluate NLP techniques and their real-world applications.

**Prerequisite**

- 23DS210: Theory of Probability
- 23DS220: Applied Statistics
- 23DS410: Linear Algebra
- 23DS540: Machine Learning
- 23DS610: Deep Learning

**Course Outcomes**

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

CO1	Explain the fundamentals of Natural Language Processing, including linguistic levels, computational challenges, FSA and FST.	Understand
CO2	Apply lexical processing techniques and syntactic analysis techniques using appropriate algorithms.	Apply
CO3	Apply semantic processing techniques including word embeddings, word sense disambiguation, and semantic role labelling for language understanding.	Apply
CO4	Apply discourse and pragmatic aspects of language, including coreference resolution, discourse relations, and sentiment interpretation.	Apply
CO5	Apply deep learning models for NLP, including RNNs, attention mechanisms, Transformers, and adaptation techniques.	Apply
CO6	Analyze NLP models for real-world applications such as question answering, text summarization, NLI, machine translation, and RAG.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	L	-	-	-	-	-	-	-	-	-	L	L	L
CO2	M	M	-	L	M	-	-	-	-	-	-	L	M	M
CO3	M	M	-	L	M	-	-	-	-	-	-	L	M	M
CO4	M	M	-	L	L	-	-	-	-	-	-	L	M	M
CO5	M	M	-	M	M	-	-	-	-	-	-	M	S	M
CO6	M	M	M	M	M	L	-	-	-	-	-	M	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal											
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6						
CO1	5	10																													4	5				



3. Christopher D. Manning and Hinrich Schütze, "Foundations of Statistical Natural Language Processing", MIT Press, 1<sup>st</sup> edition, 2009.

### Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>Introduction to NLP</b>	
1.1	What is NLP and why is it computationally hard	1
1.2	Levels of language processing (Phonological, morphological, lexical, syntactic, semantic, pragmatic)	1
1.3	History of NLP (rule-based, statistical, neural era)	1
1.4	FSA (pattern matching, morphological parsing)	1
1.5	FST (finite-state morphological analysis and generation)	1
<b>2</b>	<b>Lexical Level</b>	
2.1	Tokenization and Normalization	1
2.2	Stop word removal	1
2.3	Spelling correction: minimum edit distance and Levenshtein distance	1
2.4	Subword tokenization (Byte Pair Encoding / WordPiece)	1
2.5	Morphological analysis – Stemming	1
2.6	Morphological analysis – Lemmatization	1
2.7	POS tagging (rule-based, HMM – Viterbi algorithm, CRF)	3
2.8	NER (IOB tagging, rule-based and statistical approaches)	1
<b>3</b>	<b>Syntactic Level</b>	
3.1	CFG	1
3.2	Parse trees	1
3.3	Constituency parsing and Dependency parsing	1
3.4	Parsing algorithms: Graph-based parsing	1
3.5	Parsing algorithms: Transition-based parsing	1
<b>4</b>	<b>Semantic Level</b>	
4.1	Word meaning and WordNet	1
4.2	Word Sense Disambiguation (WSD – Lesk algorithm, knowledge-based methods)	1
4.3	Distributional semantics	1

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
4.4	BoW and TF-IDF representations	1
4.5	N-gram (unigram, bigram, trigram) and Smoothing (add-one/Laplace, Kneser-Ney)	1
4.6	Word2Vec (CBOW and Skip-gram)	1
4.7	GloVe and FastText	1
4.8	Sentence embeddings (Sentence-BERT)	1
4.9	Semantic Role Labelling	1
<b>5</b>	<b>Discourse and Pragmatic Level</b>	
5.1	Discourse coherence	1
5.2	Discourse Processing: coreference resolution	1
5.3	Discourse Processing: anaphora resolution	1
5.4	Discourse connectives	1
5.5	Speech acts	1
5.6	Grice's maxims	1
5.7	Sentiment analysis (Naive Bayes/ Logistic Regression)	1
<b>6</b>	<b>Advanced Models</b>	
6.1	RNN (LSTM/ GRU)	1
6.2	Sequence-to-Sequence model: encoder-decoder architecture and teacher forcing	1
6.3	Attention Mechanism	1
6.4	Transformer architecture (self-attention, positional encoding, BERT, GPT)	2
6.5	LLM (in-context learning, RLHF; Adaptation techniques: LORA, QLORA)	2
<b>7</b>	<b>Applications of NLP</b>	
7.1	Question Answering	1
7.2	Text Summarization (extractive and abstractive; BLEU/ROUGE metrics)	1
7.3	Natural Language Inference (NLI)	1
7.4	Machine Translation (seq2seq with attention; overview of Neural MT)	1
7.5	Retrieval Augmented Generation (RAG)	1
	Total	48

**Course Designer(s):**

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**23DS930****COMPUTER VISION**

Category	L	T	P	Credit
PCC	3	1	0	4

**Preamble**

Computer Vision is a core programme course that deals with enabling machines to interpret and understand visual information from the world. This course covers the complete perception pipeline – from image formation and low-level processing to feature extraction, segmentation, pattern analysis, and object recognition. Students will gain both theoretical foundations and practical skills applicable to real-world vision systems.

**Prerequisite**

- 23DS410 : Linear Algebra
- 23DS440 : Advanced Data Structures
- 23DS610 : Deep Learning

**Course Outcomes**

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

CO1	Explain the principles of image formation, geometric transformations and Fourier analysis, and execute filtering and enhancement operations on digital images	Understand
CO2	Detect image features using edge detectors (Canny, LoG, DoG), corner detectors (Harris, Hessian-Affine), image pyramids and Gabor-based texture descriptors	Apply
CO3	Develop image segmentation solutions including region growing, graph-cut, mean-shift, MRF-based and texture segmentation approaches	Apply
CO4	Construct object detection and recognition systems using structural, model-based, appearance-based and discriminative part-based paradigms including Bag of Words and Implicit Shape Models	Apply
CO5	Select appropriate pattern analysis methods — clustering, classification and dimensionality reduction — for given computer vision problems	Apply
CO6	Develop computer vision systems to be deployed in real-world domains such as CBIR, biometrics and activity recognition.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	M	L	M	-	-	-	-	-	-	M	S	M
CO2	S	S	S	M	M	-	-	-	-	-	-	M	S	M
CO3	S	S	S	S	M	-	-	-	-	-	-	M	S	S
CO4	S	S	S	S	M	-	-	-	L	L	-	M	S	S
CO5	S	S	S	M	M	-	-	-	-	-	-	M	S	S
CO6	M	S	S	M	S	M	L	-	M	M	-	S	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	10	10							40																4	5				
CO2	10	10	20						30																4	5	10			
CO3	10	10	20						30																4	5	10			
CO4													10	10	10						30			4	5	10				
CO5													5	10	20						30			2	5	10				
CO6													5	10	20						40			2	5	10				
Total (in %)	100						100						100						100						100					

**Syllabus**

**Image Formation and Low-Level Processing** – Image Formation: Camera Models, Sensing – Transformations: Orthogonal, Euclidean, Affine, Projective – Fourier Transform – Convolution and Filtering – Image Enhancement and Restoration – Histogram Processing

**Feature Extraction** – Edge Detection: Canny, Log, Dog – Line Detection: Hough Transform – Corner Detection: Harris, Hessian-Affine – Image Pyramids and Scale Space – Gaussian Derivative Filters – Orientation Histograms – Texture Descriptors: Gabor Filters

**Image Segmentation** – Region Growing – Edge-Based Segmentation – Graph-Cut Segmentation – Mean-Shift Clustering – Markov Random Fields – Texture Segmentation

**Object Detection and Recognition** – Object Detection: Sliding Window, Region Proposals – Recognition Methods: Structural, Model-Based, Appearance and Shape-Based – Probabilistic Paradigms – Discriminative Part-Based Models – Bag of Words (BOW) – Implicit Shape Model (ISM)

**Pattern Analysis and Learning** – Clustering: K-Means, K-Medoids, Mixture of Gaussians – Classification: Discriminant Functions, Supervised, Unsupervised, Semi-Supervised – Classifiers: Bayes, KNN, ANN – Dimensionality Reduction: PCA, LDA, ICA – Non-Parametric Methods

**Applications and Case Studies** – Content-Based Image Retrieval (CBIR) — Biometrics: Face Recognition, Fingerprint Analysis – Activity Recognition

**Reference Books & web resources**

1. Richard Szeliski, "Computer Vision: Algorithms and Applications", 2nd Edition, Springer, 2022.
2. David A. Forsyth and Jean Ponce, "Computer Vision: A Modern Approach", 2nd Edition, Pearson, 2012.
3. Richard Hartley and Andrew Zisserman, "Multiple View Geometry in Computer Vision", 2nd Edition, Cambridge University Press, 2003.

**Course Contents and Lecture Schedule**

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>1</b>	<b>Image Formation and Low-Level Processing</b>	
1.1	Image Formation: Camera Models, Pinhole Camera, Sensing, Digitisation	2
1.2	Geometric Transformations: Orthogonal, Euclidean, Affine, Projective	2
1.3	Fourier Transform, Convolution and Filtering	2
1.4	Image Enhancement: Histogram Processing, Restoration	2
<b>2</b>	<b>Feature Extraction</b>	
2.1	Edge Detection: Canny, LoG (Laplacian of Gaussian), DoG (Difference of Gaussian)	3
2.2	Line Detection: Hough Transform; Corner Detection: Harris, Hessian-Affine	3
2.3	Image Pyramids, Scale Space, Gaussian Derivative Filters	3
2.4	Orientation Histograms and Texture Descriptors: Gabor Filters	2
<b>3</b>	<b>Image Segmentation</b>	
3.1	Region Growing and Edge-Based Segmentation Approaches	2
3.2	Graph-Cut Segmentation and Markov Random Fields (MRFs)	2
3.3	Mean-Shift Clustering and Texture Segmentation	3
<b>4</b>	<b>Object Detection and Recognition</b>	
4.1	Object Detection Frameworks: Sliding Window and Region Proposals	2
4.2	Recognition: Structural, Model-based, Appearance and Shape-based Methods	2
4.3	Probabilistic Paradigms and Discriminative Part-based Models	2
4.4	Bag of Words (BOW) and Implicit Shape Model (ISM)	2
<b>5</b>	<b>Pattern Analysis and Learning</b>	
5.1	Clustering: K-Means, K-Medoids, Mixture of Gaussians (EM Algorithm)	2
5.2	Classification: Bayes, KNN, ANN; Supervised, Unsupervised, Semi-supervised	3

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
5.3	Dimensionality Reduction: PCA, LDA, ICA; Non-parametric Methods	3
<b>6</b>	<b>Applications and Case Studies</b>	
6.1	Content-Based Image Retrieval (CBIR)	2
6.2	Biometrics: Face Recognition, Fingerprint Analysis;	2
6.3	Activity Recognition	2
	<b>Total</b>	<b>48</b>

**Course Designer:**

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**23DS970****WEB ANALYTICS LAB**

Category	L	T	P	Credit
PCC	0	0	4	2

**Preamble**

The aim of the course is to enable the students to different web analytics tools and implement the data collection, acquisition and processing in terms of web analytics.

**Prerequisite**

- 23DS580 –Web Technology Lab
- 23DS370 - Applied Statistics and Python Programming Lab

**Course Outcomes**

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

CO1	Perform data collection using web analytics tools and tracking techniques.	Apply
CO2	Implement web log analysis and interpret basic website metrics.	Apply
CO3	Apply web analytics metrics to measure reach, acquisition, and conversion.	Apply
CO4	Generate analytical reports and dashboards using analytics tools.	Apply
CO5	Analyze user behavior, retention, and customer journey using analytics data.	Analyse
CO6	Implement Google Analytics	Apply

**Mapping with Programme Outcomes**

Cos	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	M	-	-	S	-	-	-	-	-	-	-	M	S
CO2	M	S	-	M	M	-	-	-	-	-	-	-	M	M
CO3	M	S	M	-	M	-	-	-	-	-	-	-	M	M
CO4	-	M	M	-	S	-	-	-	L	M	-	-	M	S
CO5	-	S	M	-	M	-	-	-	-	-	-	-	M	S
CO6	-	S	M	-	S	-	-	-	-	-	-	-	M	S

S- Strong; M-Medium; L-Low

**List of Experiments**

1. Data collection using different analytics methods (log files & page tagging).
2. Implementation of web log analysis and traffic measurement.
3. Tracking user sessions, users, and page views using analytics tools.
4. Identifying reach and audience metrics.
5. Measuring acquisition channels
6. Site visit survey reporting
7. Conversion tracking
8. Retention and cohort analysis.
9. Report generation and dashboard creation.
10. Implementation of Google Analytics.
11. Mini project involving collection, analysis, visualization, and interpretation of website or digital marketing data using web analytics tools and dashboards.

**Reference Books & web resources**

1. Bernard J. Jansen, *Understanding User-Web Interactions via Web Analytics*, 1st Edition, Springer Cham, 2022
2. Avinash Kaushik, *Web Analytics2.0*, John Wiley and Sons, 2010.

3. Web Analytics: An Hour a Day, Avinash Kaushik,2007
4. Brian Clifton, Advanced web metrics with Google analytics, John Wiley and Sons,2012.

**Course Designer(s):**

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**23DS980****NATURAL LANGUAGE  
PROCESSING LAB**Category L T P Credit  
PC 4 2**Preamble**

This laboratory course aims to provide hands-on implementation experience in Natural Language Processing, enabling students to build text processing pipelines, implement classical NLP algorithms, train word embedding models, and develop lightweight NLP application systems using Python-based tools and libraries.

**Prerequisite**

- 23DS270: Python Programming and Applied Statistics Lab,
- 23DS540: Machine Learning,
- 23DS610: Deep Learning

**Course Outcomes**

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

<b>CO1</b>	Build a text preprocessing pipeline by implementing tokenization, stop word removal, stemming, and lemmatization.	Apply
<b>CO2</b>	Implement POS tagging using HMM-based taggers and perform named entity recognition using spaCy to extract structured information from text.	Apply
<b>CO3</b>	Construct syntactic parse trees and dependency structures using Transition based Parser and Graph based parser	Apply
<b>CO4</b>	Train Word2Vec and GloVe embedding models on custom corpora, generate Sentence-BERT embeddings, implement N-gram language models, and perform word sense disambiguation using the Lesk algorithm and WordNet similarity measures.	Apply
<b>CO5</b>	Develop sentiment analysis systems using machine learning classifiers (Naive Bayes, Logistic Regression) and sequence processing using LSTM, GRU, and BiLSTM models for text classification.	Apply
<b>CO6</b>	Build lightweight NLP applications for extractive text summarization and rule-based question answering, and utilize HuggingFace Transformers pipelines including BPE tokenization exploration and zero-shot classification.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
<b>CO1</b>	M	M	S	-	S	-	-	-	-	-	-	M	S	S
<b>CO2</b>	M	M	S	M	S	-	-	-	-	-	-	M	S	S
<b>CO3</b>	M	M	S	M	S	-	-	-	-	-	-	M	S	M
<b>CO4</b>	S	M	S	S	S	-	-	-	-	-	-	S	S	S
<b>CO5</b>	S	S	S	S	S	M	-	M	M	M	-	S	S	S
<b>CO6</b>	M	S	S	S	S	M	-	M	M	M	M	S	S	S

S- Strong; M-Medium; L-Low

**List of Experiments****1. Text Processing and Preprocessing**

Tokenization - use regex for extracting from raw text - Stop Word Removal - Spelling correction - Subword tokenization - Stemming - Lemmatization

**2. Part-of-speech tagging & Named Entity**

Brill & HMM – NER

**3. Syntactic and Parsing**

Transition based Parser – Graph based parser

4. **Semantics**

N-gram language modelling (unigram, bigram, trigram) - smoothing - Word2Vec – GloVe - Sentence-BERT embeddings - BPE tokenization - WordNet - Word Sense Disambiguation - Embedding Visualization

5. **Discourse –**

Coreference resolution

6. **Sentiment Analysis –**

Naive Bayes – LR

7. **Sequence Processing –**

LSTM - GRU - BiLSTM (text classification)

8. **Transformer Pipelines –**

HuggingFace Transformers pipeline (sentiment analysis, zero-shot classification) – BPE tokenization exploration

9. **Applications –**

Extractive Text Summarization - Rule-based Question Answering

**Mini Project:**

1. BERT fine-tuning for text classification  
(or)  
GPT-based text generation with decoding strategies;
2. LoRA/QLoRA fine-tuning using Hugging Face PEFT  
(or)  
End-to-End RAG pipeline using FAISS and an LLM.

**Reference Books & web resources**

1. Daniel Jurafsky and James H. Martin, "Speech and Language Processing: An Introduction to Natural Language Processing, Computational Linguistics and Speech Recognition", Prentice Hall, 3<sup>rd</sup> Edition, 2026
2. Lewis Tunstall, Leandro von Werra, and Thomas Wolf, "Natural Language Processing with Transformers: Building Language Applications with Hugging Face", O'Reilly Media, 1<sup>st</sup> edition, 2022.
3. Sowmya Vajjala, Bodhisattwa Majumder, Anuj Gupta, Harshit Surana, "Practical natural language processing: a comprehensive guide to building real-world NLP systems", O'Reilly Media, 1<sup>st</sup> edition, 2020.
4. Steven Bird, Ewan Klein, Edward Loper, "Natural language processing with Python: analyzing text with the natural language toolkit", O'Reilly Media, Inc., 1<sup>st</sup> edition, 2009.
5. Christopher D. Manning and Hinrich Schütze, "Foundations of Statistical Natural Language Processing", MIT Press, 1<sup>st</sup> edition, 2009.

**Course Designer(s):**

1. Ms. Srimathi S, ssids@tce.edu  
Assistant Professor,  
Dept. of Applied Mathematics and Computational Science

**23DSP20****PROJECT WORK II**

Category	L	T	P	Credit
EEC	0	0	12	12

**Preamble**

This course aims to enhance the professional competencies of M.Sc. Data Science students by enabling them to design, analyze, and develop data-driven solutions. It emphasizes defining problem objectives, selecting appropriate models, and evaluating results through practical or research-based approaches. The project provides hands-on experience in applying theoretical concepts to real-world problems, improving analytical, programming, and problem-solving skills. It also promotes teamwork, project management, and communication abilities, while preparing students for careers in industry, research, and interdisciplinary domains.

**Prerequisite**

- NIL

**Course Outcomes**

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

CO1	Formulate real-world data science problems by identifying relevant concepts, functions, and system architectures for the selected domain.	Analyze
CO2	Assess and choose appropriate design alternatives by considering practical limitations, ethical aspects, and data governance policies.	Evaluate
CO3	Utilize advanced statistical, computational, and machine learning methods with suitable tools to develop effective solutions.	Create
CO4	Analyze and validate experimental or analytical results to extract meaningful insights and draw logical conclusions.	Evaluate
CO5	Exhibit teamwork, leadership, and effective communication skills in industrial, societal, or research-oriented settings.	Apply
CO6	Develop, document, and present technical reports and research outcomes in accordance with national academic standards.	Create

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	L	M	M	S	M	-	-	-	-	-	-	S	L	M
CO2	L	M	L	M	M	-	-	-	-	-	-	S	L	M
CO3	L	M	L	M	L	-	-	-	M	M	-	S	L	L
CO4	L	M	L	M	L	-	-	-	M	M	-	S	L	L
CO5	L	M	L	M	L	-	-	-	M	M	-	S	L	L
CO6	L	M	L	L	L	-	M	-	L	L	L	L	L	L

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	Review 1						Review 2						Review 3						Terminal (Viva Voce)					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																								
CO1				5																		5		
CO2					5																		5	
CO3											10												5	
CO4												10												5
CO5															15								10	
CO6																	15							10
Total (in %)	10						20						30						40					

**Project Requirements - First Review*****Expectations from Students (in the Presentation)***

<b>Review-I</b>	Title
	About the organization
	Abstract
	Introduction
	Proposed System
	Modules Split-up and Gantt Chart
	Architectural Design for Proposed System
	ER Diagram, DFD, Use case diagram ( if necessary)
	Algorithms / Techniques used
	Expected outcomes
	References
	30% of code implementation/research works

**Project Requirements - Second Review*****Expectations from Students (in the Presentation)***

<b>Review-II</b>	Title
	Abstract
	Functionalities with description
	Risk Analysis
	Data Base Design – Table design and relationships
	Detailed Design –UML Diagrams
	Results obtained (intermediate)- Screenshots
	Unit Testing screen shots for modules completed so far
	Integration Testing screen shots
	References
	70% of code implementation/research works

**Project Requirements – Third Review*****Expectations from Students (in the Presentation)***

<b>Review-III</b>	Title
	Abstract
	Overall Design
	Implementation of all Modules-Screenshots
	Experimental Results
	Performance Evaluation
	Comparison with Existing system
	Unit Testing screen shots for all modules
	Integration Testing screen shots
	References
	100% of code implementation/ research works

**Course Designer:**

Dr. T.Chandrakumar  
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**23DSPA0****HIGH PERFORMANCE  
COMPUTING**Category L T P Credit  
PEC 3 0 0 3**Preamble**

This course introduces students to the architecture and programming techniques of high-performance computing (HPC) systems, focusing on parallelism across shared and distributed memory systems.

**Prerequisite**

- 21DS130: Computer Organization
- 21DS350: Operating Systems

**Course Outcomes**

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

CO1	Understanding the fundamentals of computer organization, including processor design, memory hierarchies, and system-level hardware architecture.	Understand
CO2	Apply advanced computational algorithms in a high-performance computing environment using shared memory and message-passing paradigms.	Apply
CO3	Analyze the efficiency of parallel algorithms, focusing on memory access patterns, data locality, and synchronization overhead.	Analyze
CO4	Apply parallel programs for multicore and GPU architectures utilizing OpenMP, MPI, and CUDA programming models.	Apply
CO5	Apply performance tuning techniques to optimize resource usage across different parallel architectures.	Apply
CO6	Analyze the trade-offs between different HPC architectures (such as CPU, GPU, and accelerators) in terms of power, scalability, and performance.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	L	L	L	L	L	-	-	-	-	-	-	M	L
CO2	L	M	S	M	M	M	-	-	-	-	-	-	S	L
CO3	L	S	M	S	M	M	-	-	-	-	-	-	S	L
CO4	L	M	S	M	S	M	-	-	-	-	-	-	S	L
CO5	L	M	M	S	S	M	-	-	-	-	-	-	S	L
CO6	L	S	M	M	M	S	-	-	-	-	-	-	S	L

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT 1						Assignment 1						CAT 2						Assignment 2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	5	10																							3	10				
CO2		10	30						50																3	3	10			
CO3	5	10		30						50															3	3		15		
CO4													5	10	10					25				3	3	10				
CO5														10	20					25				3	3	10				
CO6													5	10		30					50				3		15			
<b>Total (in %)</b>	100						100						100						100											

## Syllabus

**INTRODUCTION TO HIGH PERFORMANCE COMPUTING (HPC):** Definition of High-Performance Computing (HPC) and Need and scope of HPC in modern computing, Historical perspective and evolution of HPC systems - Moore's Law and its relevance to HPC Challenges of scaling computing performance - Types of HPC systems: Supercomputers, Clusters, Cloud-based HPC - Overview of high-performance hardware architectures - Processor topology and interconnection networks - Role of storage, power supply, and cooling in HPC design - HPC case studies: Applications in weather forecasting, molecular dynamics, and big data analytics - Demonstration of a basic HPC problem

**PARALLEL PROGRAMMING BASICS:** Shared memory vs distributed memory systems: Architectural differences Memory access patterns and data locality in shared memory systems - Distributed memory systems: Inter-process communication and synchronization - Introduction to parallel processing models (SPMD, SIMD) - Multicore processors: Basics and architectural implications - Cache coherence and memory consistency models in multicore systems - Parallel algorithm design principles - Decomposition techniques for parallel algorithms - Load balancing in parallel programs: Static vs dynamic load balancing - Communication overheads in parallel programs - Overview of basic parallel programming paradigms

**MESSAGE PASSING INTERFACE (MPI):** Introduction to the Message Passing Interface (MPI) standard - Concepts of message passing and process communication - Basic MPI functions: MPI\_Init, MPI\_Send, MPI\_Recv, MPI\_Finalize - Point-to-point communication and blocking vs non-blocking operations - Collective communication operations: Broadcast, Scatter, Gather, Reduce - Groups and communicators in MPI - Synchronization in MPI: Barriers and synchronization mechanisms - Error handling and debugging in MPI programs - Domain decomposition methods for scientific computing

**OPENMP:** Introduction to OpenMP: Fork-Join parallelism model - Basics of OpenMP directives for parallel regions - Work-sharing constructs in OpenMP: #pragma omp for, #pragma omp sections - Scheduling and controlling parallel loops - Synchronization in OpenMP: Critical sections, barriers, and atomic operations - Handling race conditions in OpenMP programs - OpenMP reductions: Using reduction clauses for aggregation operations - Thread-private variables and memory consistency in OpenMP-Nested parallelism in OpenMP

**GPU COMPUTING WITH CUDA:** Introduction to GPU architecture and CUDA programming - Differences between CPU and GPU processing models - CUDA programming model: Thread blocks, warps, and grid structures - Writing basic CUDA kernels for parallel computation - GPU memory hierarchy: Global, shared, and local memory - Managing memory and optimizing memory access patterns in CUDA - Synchronization and parallel reduction in CUDA - Using CUDA libraries for matrix algebra (e.g., cuBLAS)

**PERFORMANCE OPTIMIZATION:** Introduction to performance profiling tools (e.g., gprof, Valgrind, NVProf) - Measuring and analyzing the performance of parallel programs - Scalability metrics: Speedup, efficiency, and Amdahl's Law - Strong and weak scaling in parallel systems - Minimizing communication overheads in parallel programs - Techniques to reduce latency and improve data locality - Cache optimization techniques: Tiling, blocking, and loop unrolling

## Reference Books & web resources

1. L. Hennessy and D. A. Patterson, "Computer Architecture: A Quantitative Approach", Morgan Kaufmann, 2017.

2. M. Quinn, "Parallel Programming in C with MPI and OpenMP", McGraw-Hill Education, 2017.
3. A. Grama, et al., "Introduction to Parallel Computing", Pearson, 2018.
4. G. Zaccane, "Python Parallel Programming Cookbook", Packt Publishing, 2017.
5. [OpenMP Documentation](https://www.openmp.org)
6. [MPI Standard](https://www.mpi-forum.org)
7. [CUDA Programming Guide](https://developer.nvidia.com/cuda-toolkit)

### Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>INTRODUCTION TO HPC</b>	
1.1	Definition of High-Performance Computing (HPC) and Need and scope of HPC in modern computing, Historical perspective and evolution of HPC systems	1
1.2	Moore's Law and its relevance to HPC - Challenges of scaling computing performance - Types of HPC systems: Supercomputers, Clusters, Cloud-based HPC	2
1.3	Overview of high-performance hardware architectures - Processor topology and interconnection networks - Role of storage, power supply, and cooling in HPC design	1
1.4	HPC case studies: Applications in weather forecasting, molecular dynamics, and big data analytics - Demonstration of a basic HPC problem (e.g., matrix multiplication)	1
2	<b>PARALLEL PROGRAMMING BASICS</b>	
2.1	Shared memory vs distributed memory systems: Architectural differences - Memory access patterns and data locality in shared memory systems	1
2.2	Distributed memory systems: Inter-process communication and synchronization - Introduction to parallel processing models (SPMD, SIMD)	1
2.3	Multicore processors: Basics and architectural implications - Cache coherence and memory consistency models in multicore systems	1
2.4	Parallel algorithm design principles - Decomposition techniques for parallel algorithms	1
2.5	Load balancing in parallel programs: Static vs dynamic load balancing	1
2.6	Communication overheads in parallel programs	1
2.6	Overview of basic parallel programming paradigms	1
3	<b>MESSAGE PASSING INTERFACE (MPI)</b>	
3.1	Introduction to the Message Passing Interface (MPI) standard Concepts of message passing and process communication	1
3.2	Basic MPI functions: MPI_Init, MPI_Send, MPI_Recv, MPI_Finalize Point-to-point communication and blocking vs non-blocking operations	1
3.3	Collective communication operations: Broadcast, Scatter, Gather, Reduce - Groups and communicators in MPI	1
3.4	Synchronization in MPI: Barriers and synchronization mechanisms Error handling and debugging in MPI programs	1
3.5	Domain decomposition methods for scientific computing - Case study: Parallelizing a numerical method (e.g., solving linear equations) using MPI	1
3.6	Performance considerations for MPI programs - Optimizing communication and computation overlap in MPI applications	1

4	<b>OPENMP</b>	
4.1	Introduction to OpenMP: Fork-Join parallelism model - Basics of OpenMP directives for parallel regions	1
4.2	Work-sharing constructs in OpenMP: #pragma omp for, #pragma omp sections - Scheduling and controlling parallel loops	1
4.3	Synchronization in OpenMP: Critical sections, barriers, and atomic operations - Handling race conditions in OpenMP programs	1
4.4	OpenMP reductions: Using reduction clauses for aggregation operations - Thread-private variables and memory consistency in OpenMP	1
4.5	Nested parallelism in OpenMP - Case study: Parallelizing matrix multiplication with OpenMP	1
4.6	Performance tuning for OpenMP programs - Scalability and bottlenecks in shared memory parallel applications	1
5	<b>GPU COMPUTING WITH CUDA</b>	
5.1	Introduction to GPU architecture and CUDA programming - Differences between CPU and GPU processing models - CUDA programming model: Thread blocks, warps, and grid structures - Writing basic CUDA kernels for parallel computation	2
5.2	GPU memory hierarchy: Global, shared, and local memory - Managing memory and optimizing memory access patterns in CUDA	1
5.3	Synchronization and parallel reduction in CUDA - Using CUDA libraries for matrix algebra (e.g., cuBLAS)	1
5.4	Case study: Matrix multiplication on a GPU using CUDA - Optimizing the performance of CUDA programs	1
5.5	Performance tuning for GPU applications: Profiling tools and common optimization techniques - Case study: Performance comparison of CPU vs GPU for a given problem	1
6	<b>PERFORMANCE OPTIMIZATION</b>	
6.1	Introduction to performance profiling tools (e.g., gprof, Valgrind, NVProf) - Measuring and analyzing the performance of parallel programs	1
6.2	Scalability metrics: Speedup, efficiency, and Amdahl's Law - Strong and weak scaling in parallel systems	1
6.3	Minimizing communication overheads in parallel programs - Techniques to reduce latency and improve data locality	1
6.4	Cache optimization techniques: Tiling, blocking, and loop unrolling - Case study: Optimizing a memory-bound application	1
6.5	Power and energy optimization in HPC systems - Green computing and its importance in HPC environments	1
6.6	Case study: Performance tuning of a complex scientific application Comparison of performance metrics across CPU, MPI, OpenMP, and CUDA implementations	2
	<b>Total</b>	<b>36</b>

**Course Designer:**

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**23DSPB0****MOBILE APPLICATION  
DEVELOPMENT**

Category L T P Credit

PEC 3 0 0 3

**Preamble**

This course provides a comprehensive introduction to mobile app development, focusing on Android architecture, UI design, data storage, and cross-platform tools like PhoneGap and HTML5.

**Prerequisite**

- 23DS530 - Web Technology
- 23DS480 - Java Programming Lab

**Course Outcomes**

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

CO1	Explain the types, features, and challenges of mobile application development frameworks.	Understand
CO2	Describe Android features, architecture, and development environment, including installation and configuration processes.	Understand
CO3	Compare and contrast various Android APIs, focusing on their functionality and application in mobile app development.	Apply
CO4	Explain various user interface components and activities to create real-time mobile applications with efficient layouts and controls.	Apply
CO5	Implement services, data storage solutions, and location-based services for specific mobile app scenarios using SQLite, files, and content providers.	Apply
CO6	Analyze the practical applications of mobile applications in real-world scenarios, including cross-platform tools and multimedia integration.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	L	L	M	L	L	L	L	L	L	M	M	L
CO2	S	M	M	L	S	L	L	L	L	L	L	M	M	M
CO3	M	M	S	M	S	L	L	L	L	L	L	M	S	M
CO4	M	M	S	M	S	L	L	L	M	M	L	M	M	S
CO5	M	S	S	M	S	L	L	L	M	M	L	M	M	S
CO6	M	S	S	S	S	L	L	L	M	M	L	M	M	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	10	10																							5	10				
CO2	10	10																							5	5				
CO3		10	20						50																5	5	10			
CO4			30						50				10												5	5	10			
CO5													10	10	30											5	10			
CO6														10	20	10											10	10		
Total (in %)	100						100						100						100											

## Syllabus

**MOBILE APPLICATION DEVELOPMENT FRAMEWORK:** Types- Features- challenge. Android: Introduction – Architecture -Installation and configuration- ADE- APIs

**USER INTERFACES:** Components-views-Layouts and its types- Menus and its types – Dialogues and its types - Notifications and its types

**DATA STORAGE:** SQLite database-Files- shared preferences-Content Providers

**LOCATION BASED SERVICES:** Location Provider- Geo Coder- Map Multimedia: AudioVideo- Animations- Drawing.

**PHONE GAP AND HTML5:** Phone Gap Introduction – Architecture- Installation and configuration, HTML5 Introduction -Data list control- Validation feature -Application Cache in HTML 5.

**EVENTS:** Listener and its types-handler and its types. Geo location, Media, Storage.

## Reference Books & web resources

1. Lauren Darcey and Shane Conder, "Android Wireless Application Development", Pearson Education, 3rd edition. (2014) (Chapters 1 to 4).
2. RetoMeier, "Professional Android Application Development", Wrox, 2010.
3. Thomas Myer, "Beginning PhoneGap,Wrox", 2012.
4. Mark Murphy, "Beginning Android,Apress", 2009.
5. Rick Rogers et.al, "Android – Application Development", O'Reilly, 2009.
6. Jochen Schiller, "Mobile Communications", Addison Wesley, 2011.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>MOBILE APPLICATION DEVELOPMENT FRAMEWORK</b>	
1.1	Types	2
1.2	Features and challenges	1
<b>2</b>	<b>ANDROID</b>	
2.1	Introduction	1
2.2	Architecture	2
2.3	ADE	1
2.4	APIs	1
<b>3</b>	<b>USER INTERFACES</b>	
3.1	Components, views	2
3.2	Layouts and its types	2
3.3	Menus and its types	1
3.4	Dialogues and its types	1
3.5	Notifications and its types	1

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>4</b>	<b>DATA STORAGE</b>	
4.1	SQLite database, files	2
4.2	Shared preferences, Content Providers	2
<b>5</b>	<b>LOCATION BASED SERVICES</b>	
5.1	Location Provider	1
5.2	Geo coder	1
5.3	Map	1
<b>6</b>	<b>MULTIMEDIA</b>	
6.1	Audio	1
6.2	Video	1
6.3	Animations	1
6.4	Drawing	1
<b>7</b>	<b>PHONEGAP AND HTML5</b>	
7.1	PhoneGap Introduction	1
7.2	Architecture	2
7.3	HTML5 Introduction	1
7.4	Data list control	1
7.5	Validation feature	1
7.6	Application Cache in HTML 5	1
<b>8</b>	<b>EVENTS</b>	
8.1	Listener and its types	2
8.2	Handler and its types	1
	<b>Total</b>	<b>36</b>

**Course Designer(s):**

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**23DSPC0 HEALTHCARE INFORMATICS**

Category L T P Credit  
PEC 3 0 0 3

**Preamble**

This course provides an overview of healthcare informatics, focusing on the critical aspects of healthcare data, health information systems, and emerging technologies. The course aims to equip students with a practical understanding of how data science and IT support modern healthcare, and healthcare data privacy.

**Prerequisite**

Nil

**Course Outcomes**

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

CO1	Explain the role of healthcare informatics and health information systems in healthcare, and how they contribute to modern healthcare reform	Apply
CO2	Describe healthcare data, its sources, quality aspects, and the standards used to manage and share healthcare information	Apply
CO3	Utilize health information systems and data analytics to improve population health management, Tele-health, and patient engagement	Apply
CO4	Implement system acquisition, development, and support processes in a healthcare organization, considering the entire lifecycle	Apply
CO5	Use legal, privacy, and security principles to protect healthcare information systems and ensure data confidentiality	Apply
CO6	Assess emerging technologies such as AI, IoT, and Big Data, and their implications on healthcare IT governance and strategic planning	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	L	-	-	M	L	-	S	M	L	-	M	S
CO2	S	S	M	L	M	S	-	-	S	L	S	M	M	S
CO3	S	L	S	M	S	L	-	M	S	L	S	S	S	S
CO4	S	S	L	S	M	S	-	M	S	L	M	L	S	M
CO5	S	S	L	S	M	S	-	M	S	L	M	L	S	M
CO6	M	M	M	M	S	S	M	M	L	M	M	S	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	10	10	10						30																5		10			
CO2	10	10	20						30																5		10			
CO3		10	20						40																5	5	10			
CO4													10		20						30				5	5	10			
CO5													10	10	20						30						5	10		
CO6														10	20						40						5	10		
Total (in %)	100						100						100						100											

## Syllabus

**INTRODUCTION TO HEALTHCARE INFORMATICS AND THE NATIONAL HEALTH IT LANDSCAPE-** Introduction to Healthcare Informatics, National Health Information Technology Landscape, Electronic Health Records (EHR), Personal Health Records (PHR)

**HEALTHCARE DATA: SOURCES, QUALITY, AND STANDARDS-** Healthcare Data and Information

Data Sources: Clinical, Administrative, and External, Data Quality Dimensions, Healthcare Standards (HL7, ICD-10, SNOMED)

**INFORMATION SYSTEMS TO SUPPORT POPULATION HEALTH MANAGEMENT-** Population Health Management (PHM), Data analytics, dashboards, Tele health, Patient engagement in PHM

**SYSTEM ACQUISITION, IMPLEMENTATION, AND SUPPORT IN HEALTHCARE IT-** System Acquisition Process, Systems Development Life Cycle (SDLC), Post-implementation support, Case Study: Selecting an EHR for Dermatology Practice Case Study: Planning an EHR Implementation

**PRIVACY, SECURITY, AND LEGAL ASPECTS IN HEALTHCARE IT-** Privacy and Confidentiality, Legal frameworks: HIPAA, HITECH Act, Cybersecurity measures and risk management, Case Study: Breaching the Security of an Internet Patient Portal, Case Study: The Decision to Develop an IT Strategic Plan

**EMERGING TRENDS AND IT GOVERNANCE IN HEALTHCARE-** AI, IoT, and Big Data in Healthcare IT Governance and Strategic Planning, Case Study: Implementing Tele-psychiatry in a Community Hospital Emergency Department, Case Study: Assessing the Value and Impact of CPOE (Computerized Physician Order Entry)

## Reference Books & web resources

1. Wager, K. A., Lee, F. W., & Glaser, J. P. (2017). "Health Care Information Systems: A Practical Approach for Health Care Management" (4th ed.). Jossey-Bass. (Chapters 1-6, 9,12-14).
2. Shortliffe, E. H., & Cimino, J. J. (2014). "Biomedical Informatics: Computer Applications in Health Care and Biomedicine" (4th ed.). Springer.
3. Hoyt, R. E., & Yoshihashi, A. (2017). "Health Informatics: Practical Guide for Healthcare and Information Technology Professionals (7th ed.)". Informatics Education.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>INTRODUCTION TO HEALTHCARE INFORMATICS AND THE NATIONAL HEALTH IT LANDSCAPE</b>	
1.1	Introduction to Healthcare Informatics	2
1.2	National Health Information Technology Landscape	1
1.3	Electronic Health Records (EHR)	1
1.4	Personal Health Records (PHR)	1

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>2</b>	<b>HEALTHCARE DATA: SOURCES, QUALITY, AND STANDARDS</b>	
2.1	Healthcare Data and Information	1
2.2	Data Sources: Clinical, Administrative, and External	1
2.3	Data Quality Dimensions	1
2.4	Healthcare Standards (HL7, ICD-10, SNOMED)	1
<b>3</b>	<b>Information Systems To Support Population Health Management</b>	
3.1	Population Health Management (PHM)	1
3.2	Data analytics, dashboards	1
3.3	Tele health	1
3.4	Patient engagement in PHM	1
<b>4</b>	<b>SYSTEM ACQUISITION, IMPLEMENTATION, AND SUPPORT IN HEALTHCARE IT</b>	
4.1	System Acquisition Process, ,	2
4.2	Systems Development Life Cycle (SDLC)	2
4.3	Post-implementation support	2
4.4	Case Study: Selecting an EHR for Dermatology Practice	1
4.5	Case Study: Planning an EHR Implementation	1
<b>5</b>	<b>PRIVACY, SECURITY, AND LEGAL ASPECTS IN HEALTHCARE IT</b>	
5.1	Privacy and Confidentiality	1
5.2	Legal frameworks: HIPAA, HITECH Act	1
5.3	Cybersecurity measures and risk management	1
5.4	Case Study: Breaching the Security of an Internet Patient Portal	1
5.5	Case Study: The Decision to Develop an IT Strategic Plan	1
<b>6</b>	<b>EMERGING TRENDS AND IT GOVERNANCE IN HEALTHCARE</b>	
6.1	AI, IoT, and Big Data in Healthcare IT Governance and Strategic Planning	2
6.2	Case Study: Implementing Tele-psychiatry in a Community Hospital Emergency Department	1
6.3	Case Study: Assessing the Value and Impact of CPOE (Computerized Physician Order Entry)	1
	<b>Total</b>	<b>30</b>

**Course Designer:**

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**23DSPD0****EMBEDDED SYSTEMS**

Category	L	T	P	Credit
PEC	3	0	0	3

**Preamble**

The course provides comprehensive knowledge of embedded systems design, programming, and real-time operating systems. Students will learn microcontroller architectures, programming in C, Assembly, and Java, and use development tools like emulators and debuggers.

**Prerequisite**

- 23DS130 - Digital Electronics
- 23DS350 - Operating Systems
- 23DS480 – Java Programming Lab

**Course Outcomes**

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

CO1	Understand the need for embedded systems and their design process.	Understand
CO2	Analyze the architecture and functioning of embedded processors and microcontrollers.	Analyze
CO3	Apply programming techniques using assembly, C, and Java for embedded systems.	Apply
CO4	Analyze the role of real-time operating systems and the importance of interrupts and multithreading.	Analyze
CO5	Apply embedded software development tools, including debuggers and emulators, for real-world projects.	Apply
CO6	Analyze design issues of embedded systems through case studies like digital cameras and smart cards.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	M	M	M	-	-	-	-	-	-	L	M	L
CO2	L	S	S	S	S	-	-	-	-	-	-	L	S	L
CO3	L	M	S	M	M	-	-	-	-	-	-	L	S	L
CO4	L	M	S	S	S	-	-	-	-	-	-	L	S	L
CO5	L	M	M	M	S	-	-	-	-	-	-	L	S	L
CO6	L	M	S	S	S	-	-	-	-	-	-	S	S	L

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal										
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6					
CO1	5	20																							3	10									
CO2	5	5	10	20							50														3	3		10							
CO3	5	10	20								50														3	3	10								
CO4													5	5	10	20											25			3	3		15		
CO5													5	10	10											50				3	3	10			
CO6														5	10	20											25				3		15		
Total (in %)	100						100						100						100																

## Syllabus

**INTRODUCTION TO EMBEDDED SYSTEMS:** Need for Embedded Systems - Introduction to embedded systems, their significance, and challenges - Embedded System Design Process, Steps involved in designing embedded systems - Microprocessors and Microcontrollers, Introduction to embedded processors: 8051, ARM, architecture and uses.

**PROGRAMMING EMBEDDED SYSTEMS:** Introduction to Assembly Language, Basic assembly programming for embedded systems - Introduction to C for Embedded Systems, C programming and its use in efficient embedded code design - Introduction to Java for Embedded Systems, Role of Java in embedded systems and its application - Efficient Code Techniques, Best practices in embedded programming using these languages.

**REAL-TIME OPERATING SYSTEMS (RTOS):** Inter-process Communication (IPC) in RTOS - Basics of IPC mechanisms in real-time systems - Interrupt Handling - Importance and implementation of interrupts: hardware and software interrupts - Multithreading Concepts in RTOS, Threads, task scheduling, and their significance in RTOS.

**HARDWARE AND COMMUNICATION PROTOCOLS:** Interface Design: Hardware and Software, Design methodologies for interfacing hardware with embedded systems - Communication Protocols: RTC Interfacing, Introduction to interfacing protocols like RTC, use in embedded designs - Programming Communication Interfaces, Hands-on programming of communication interfaces in embedded systems.

**EMBEDDED SOFTWARE DEVELOPMENT:** Layered Software Development, Understanding of FSM controllers and layered abstraction in embedded software - Device Driver Development, Concepts of developing device drivers for embedded systems - Real-Time Constraints and Programming, Methods to meet real-time constraints through efficient software design - Emulators and Debuggers in Embedded Development, Use of debugging tools and emulators in embedded software development.

**EMBEDDED SYSTEM DESIGN TECHNIQUES & CASE STUDIES:** Design Issues and Techniques in Embedded Systems - Key design issues and techniques in modern embedded systems - Case Studies: Robot, Digital Camera, Smart Cards, Real-world design case studies showcasing complete embedded system design processes.

## Reference Books & web resources

1. Wayne Wolf, "Computers as Components: Principles of Embedded Computing System Design", 3rd edition, 2016.
2. Raj Kamal, "Embedded Systems: Architecture, Programming and Design", Tata McGraw Hill, 3rd edition, 2017.
3. Frank Vahid and Tony Givargis, "Embedded System Design: A Unified Hardware/Software Introduction", John Wiley & Sons, 2016.
4. [NPTEL - Embedded Systems] (<https://nptel.ac.in/courses/106/105/106105193/>)
5. [Coursera – Embedded Systems] (<https://www.coursera.org/specializations/embedded-systems>)

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>INTRODUCTION TO EMBEDDED SYSTEMS</b>	
1.1	Need for Embedded Systems, Introduction to embedded systems,	1

	their significance, and challenges	
1.2	Embedded System Design Process, Steps involved in designing embedded systems.	1
1.3	Microprocessors and Microcontrollers, Introduction to embedded processors: 8051, ARM, architecture and uses.	2
2	<b>PROGRAMMING EMBEDDED SYSTEMS</b>	
2.1	Introduction to Assembly Language, Basic assembly programming for embedded systems	2
2.2	Introduction to C for Embedded Systems, C programming and its use in efficient embedded code design.	2
2.3	Introduction to Java for Embedded Systems, Role of Java in embedded systems and its application.	2
2.4	Efficient Code Techniques, Best practices in embedded programming using these languages	2
3	<b>REAL-TIME OPERATING SYSTEMS (RTOS)</b>	
3.1	Inter-process Communication (IPC) in RTOS, Basics of IPC mechanisms in real-time systems	2
3.2	Interrupt Handling, Importance and implementation of interrupts: hardware and software interrupts.	2
3.3	Multithreading Concepts in RTOS, Threads, task scheduling, and their significance in RTOS	2
4	<b>HARDWARE AND COMMUNICATION PROTOCOLS</b>	
4.1	Interface Design: Hardware and Software, Design methodologies for interfacing hardware with embedded systems	2
4.2	Communication Protocols: RTC Interfacing, Introduction to interfacing protocols like RTC, use in embedded designs.	2
4.3	Programming Communication Interfaces, Hands-on programming of communication interfaces in embedded systems.	2
5	<b>EMBEDDED SOFTWARE DEVELOPMENT</b>	
5.1	Layered Software Development, Understanding of FSM controllers and layered abstraction in embedded software	2
5.2	Device Driver Development, Concepts of developing device drivers for embedded systems	2
5.3	Real-Time Constraints and Programming, Methods to meet real-time constraints through efficient software design.	2
5.4	Emulators and Debuggers in Embedded Development, Use of debugging tools and emulators in embedded software development.	2
6	<b>EMBEDDED SYSTEM DESIGN TECHNIQUES &amp; CASE STUDIES</b>	
6.1	Design Issues and Techniques in Embedded Systems, Key design issues and techniques in modern embedded systems	2
6.2	Case Studies: Robot, Digital Camera, Smart Cards, Real-world design case studies showcasing complete embedded system design processes.	2
<b>Total</b>		<b>36</b>

**Course Designer(s):**

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**23DSPE0****MARKETING ANALYTICS**

Category L T P Credit

PEC 3 0 0 3

**Preamble**

The Marketing Analytics course outcomes are highly beneficial to students as they provide practical skills in data-driven decision-making, enabling them to understand and apply key tools and techniques to real-world marketing challenges. Students gain expertise in analyzing customer behaviours using Customer Relationship Management (CRM) systems.

**Prerequisite**

- 23DS470 - Predictive Analytics
- 23DS520 - Optimization Techniques
- 23DS830 - Business Analytics

**Course Outcomes**

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

CO1	Explain the conceptual foundation of Strategic CRM.	Understand
CO2	Analyze the evolution and growth of CRM using operational CRM tools and software.	Apply
CO3	Illustrate the impact and measuring the return on investment for social media marketing.	Apply
CO4	Apply marketing analytics models and metrics.	Apply
CO5	Implement the competitive analysis strategies and business strategies using strategic models.	Apply
CO6	Analyze product, service, and pricing analytics techniques	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	L	L	L	L	L	-	S	-	-	L	L	L	M
CO2	L	S	L	L	S	L	-	L	-	-	L	L	M	L
CO3	L	L	L	M	M	S	-	M	-	-	L	L	S	M
CO4	S	S	M	S	S	L	-	L	-	-	L	L	L	M
CO5	L	S	S	L	L	L	-	L	-	-	S	L	L	M
CO6	L	L	M	L	S	S	-	L	-	-	L	L	L	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	10	10																							5	5				
CO2	5	5	30				10	40																			10			
CO3	5	5	30				10	40																	5	5	10			
CO4													5	5	10											5	10			
CO5													10	20							50					5	10			
CO6													10	20	20							50				5	5	20		
Total (in %)	100						100						100						100											

## Syllabus

**CUSTOMER RELATIONSHIP MANAGEMENT (CRM): Conceptual Foundation:** Strategic CRM Today - why managing customers more critical than Ever - benefits of customer value management approach - Evolution and Growth of CRM - Concepts of Customer value-Strategic CRM – **Operational CRM:** Software tools and Dashboard- CRM Software and Applications- CRM in Social Media.

**MARKETING ANALYTICS:** overview of Marketing Analytics - Models and metrics- Market Insight – Market data sources, sizing, PESTLE trend analysis, and porter five forces analysis – Market segment identification and positioning.

**COMPETITIVE ANALYSIS AND BUSINESS STRATEGY:** Competitor identification, Intelligence gathering, analysis and strategy- Analytics based strategy selection, with strategic models and metrics , Forecasting, balanced scorecard, and critical success factors.

**PRODUCT, SERVICE AND PRICE ANALYTICS:** Conjoint analysis model, decision tree model, portfolio resource allocation, Pricing techniques, pricing assessment, pricing for business markets, price discrimination.

**DISTRIBUTION AND PROMOTION ANALYTICS:** Retail location selection, distribution channel evaluation, and multi-channel distribution, Promotion budget estimation and allocation, promotion metrics for traditional media and social media.

**SALES ANALYTICS:** E Commerce sales mode, sales metrics, profitability metrics and support metrics.

## Reference Books & web resources

1. V.Kumar, Werner Reinartz, "Customer Relationship Management Concept, Strategy, and Tools", Third Edition, published by John Wiley & Sons, Inc., 2006. (Chapter 1,2,3,9,15)
2. Stephan Sorger, "Marketing Analytics – Strategic Models and Metrics", Admiral Press, 2013.(Chapter 1-5,7,8,9,10,11)
3. Mark Jeffery, "Data Driven Marketing: The 15 Metrics Everyone in Marketing should know", Wiley, 2013.
4. Paul W. Farris, Neil T. Bendle, Phillip E. Pfeifer, David J. Reibstein, "Marketing Metrics: The Definitive Guide to Measuring Marketing Performance", Pearson FT press, 2012.
5. [https://onlinecourses.nptel.ac.in/noc20\\_mg30/preview](https://onlinecourses.nptel.ac.in/noc20_mg30/preview) - NPTEL, By Prof. Swagato Chatterjee | IIT Kharagpur

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>CONCEPTUAL FOUNDATION OF STRATEGIC CRM</b>	
1.1	Conceptual Foundation of Strategic CRM	1
1.2	Importance of Customer Management	1
1.3	Customer Value Management Approach benefits	1

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>2</b>	<b>EVOLUTION AND GROWTH OF CRM</b>	
2.1	Software Tools and Dashboards	2
2.2	CRM Implementation: In-house Development vs. Outsourcing	2
2.3	CRM Software Applications and Importance	2
<b>3</b>	<b>CRM IN SOCIAL MEDIA</b>	
3.1	The Social Media Landscape and Role in CRM	1
3.2	Impact of Word of Mouth on Brands	1
3.3	Measuring the ROI of Social Media Marketing	1
<b>4</b>	<b>MARKETING ANALYTICS MODELS AND METRICS</b>	
4.1	Market Data Sources and Sizing	2
4.2	PESTLE Analysis	2
4.3	Porter's Five Forces Analysis	2
4.4	Market Segment Identification and Positioning	2
4.5	Competitive Analysis: Intelligence Gathering and Strategy	1
<b>5</b>	<b>BUSINESS STRATEGY AND ANALYTICS</b>	
5.1	Analytics-based Strategy Selection	1
5.2	Forecasting and Critical Success Factors	1
5.3	Balanced Scorecard Approach	1
5.4	Design and Development in Marketing Analytics	1
<b>6</b>	<b>PRODUCT, SERVICE, AND PRICE ANALYTICS</b>	
6.1	Conjoint Analysis Model and Decision Tree Model	1
6.2	Portfolio Resource Allocation	1
6.3	Pricing Techniques and Assessment	1
6.4	Pricing for Business Markets and Price Discrimination	1
6.5	Distribution and Promotion Analytics	1
<b>7</b>	<b>SALES ANALYTICS</b>	
7.1	E-commerce Sales Model	1
7.2	Sales Metrics and Profitability Metrics	1
7.3	Support Metrics for Sales	1

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
7.4	Application of Marketing Analytics in Sales	1
	<b>Total</b>	<b>34</b>

**Course Designer(s):**

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**23DSPF0 LARGE LANGUAGE MODELS**

Category L T P Credit  
PEC 3 0 0 3

**Preamble**

This course provides an in-depth exploration of Large Language Models (LLMs), covering their architectures, training methodologies, and applications. Students will engage with transformer models, prompt engineering, and fine-tuning techniques.

**Prerequisite**

- 23DS210: Theory of Probability
- 23DS410: Linear Algebra
- 23DS540: Machine Learning

**Course Outcomes**

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

CO1	Explain the transformers and attention methods.	Understand
CO2	Describe the concept of Casual language modeling and GPT	Understand
CO3	Explore Masked Language modeling, pre-training and fine-tuning language models	Apply
CO4	Apply tokenization techniques and design a vanilla attention mechanism for long range context windows.	Apply
CO5	Implement optimization strategies, including LION, Adam, gradient clipping, and evaluate models benchmarks like MMLU and HELM.	Apply
CO6	Explore and apply fine-tuning methods like prompt tuning, multi-task fine-tuning, and evaluate with different frame works.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	S	S	S	L	-	-	-	-	-	-	L	M	M
CO2	M	S	M	S	M	-	-	-	-	-	-	L	M	S
CO3	M	M	S	S	S	-	-	-	-	-	-	L	S	M
CO4	S	S	M	S	M	L	-	-	-	-	-	L	S	S
CO5	S	S	M	S	M	L	-	-	-	-	-	L	S	S
CO6	M	S	S	M	S	-	-	-	-	-	-	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's Level																														
CO1	10	15					20																		4	6				
CO2	10	15					30																		4	6				
CO3	10	20	20					50																	4		15			
CO4													10		20					30					6	15				
CO5														10	20					40					6	15				
CO6													10	10	20					30				4		15				
Total (in %)	100						100						100						100						100					

## Syllabus

**Transformers:** Introduction to transformers - Self-attention - cross-attention-Masked attention-Positional encoding, A deep dive into number of parameters, computational complexity and FLOPs

**Language Modeling:** Causal Language Modeling: What is a language model- Generative Pretrained Transformers (GPT) - Training and inference, Masked Language Modeling. Bidirectional Encoder Representations of Transformers (BERT) -Transfer Learning and Fine-Tuning, Pre-training vs Fine-tuning.

**Tokenization and Datasets:** BPE, SentencePiece, wordpiece, Bigger Picture: T5, A deep dive into text-to-text (genesis of prompting), taxonomy of models, road ahead, **Data:** Datasets, Pipelines, effectiveness of clean data, Architecture: Types of attention, positional encoding (PE) techniques, scaling techniques

**Optimization and Training Techniques:** Training: Revisiting optimizers, LION vs Adam, Loss functions, Learning rate schedules learning rate scheduling strategies: linear warm-up, cosine annealing, Gradient Clipping, typical failures during training,.

**Fine-tuning and Evaluation**–Prompt Tuning, Multi-task Fine-tuning, Parametric Efficient Fine-Tuning, Instruction fine-tuning datasets.MMLU, BigBench, HELM, OpenLLM.

**Evaluation Frameworks:** Training Large Models: Mixed precision training, Activation checkpointing,3D parallelism, ZERO, Bloom as a case study. Scaling Laws: Chinc,hilla, Gopher, Palm v2

## Reference Books & web resource

1. Ian Goodfellow, YoshuaBenjio, Aaron Courville, “Deep Learning”, The MIT Press. 2016
2. ZakariaSabti, “Prompt Engineering Demystified: Unleashing the Power of Large Language Models”, 2023
3. Andrew Radford, “Transformational Grammar: A First Course.”
4. Stephen Wolfram, “What is ChatGPT Doing ... and Why Does It Work?”
5. <https://stanford-cs324.github.io/winter2022/lectures/>

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>TRANSFORMERS</b>	
1.1	Introduction to transformers	1
1.2	Self-attention - cross- attention	1
1.3	Masked attention- Positional encoding	1
1.4	A deep dive into number of parameters, computational complexity and FLOPs	2
<b>2</b>	<b>LANGUAGE MODELING</b>	
2.1	Causal Language Modeling: What is a language model?-	1
2.2	Generative Pretrained Transformers (GPT) - Training and inference,	1

Module No.	Topic	No. of Periods
2.3	Masked Language Modeling : Bidirectional Encoder Representations of Transformers (BERT)	1
2.4	Transfer Learning and Fine-Tuning	1
2.5	Pre-training vs Fine-tuning	1
<b>3</b>	<b>TOKENIZATION AND DATASETS</b>	
3.1	BPE, SentencePiece, wordpiece, Bigger Picture: T5,	1
3.2	A deep dive into text-to-text (genesis of prompting), taxonomy of models, road ahead	2
3.3	Data: Datasets, Pipelines, effectiveness of clean data,	1
3.4	Architecture: Types of Attention, Positional Encoding (PE) Techniques, Scaling Techniques	2
<b>4</b>	<b>OPTIMIZATION AND TRAINING TECHNIQUES</b>	
4.1	Training: Revisiting optimizers, LION vs Adam,	2
4.2	Loss functions, Learning rate schedules	1
4.3	Learning rate scheduling strategies: linear warm-up, cosine annealing,	1
4.4	Gradient Clipping, typical failures during training.	1
<b>5</b>	<b>FINE-TUNING AND EVALUATION</b>	
5.1	Prompt Tuning, Multi-task Fine-tuning	2
5.2	Parametric Efficient Fine-Tuning, Instruction fine-tuning datasets	2
<b>6</b>	<b>EVALUATION AND BENCHMARKING</b>	
6.1	MMLU, BigBench, HELM, OpenLLM	3
6.2	Evaluation Frameworks,	1
6.3	Training Large Models: Mixed precision training, Activation checkpointing,	1
6.4	3D parallelism, ZERO, Bloom as a case study	1
6.5	Scaling Laws: Chinchilla,	2
6.6	Gopher, Palm v2	2
	<b>Total</b>	<b>35</b>

**Course Designer(s):**

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**23DSPG0****SOFT COMPUTING**

Category	L	T	P	Credit
PE	3	0	0	3

**Preamble**

The course is designed to introduce students to soft computing concepts and techniques and foster their abilities in analysing soft computing-based solutions for real- world problems.

**Prerequisite**

- 23DS520 – Optimization Techniques
- 23DS540 – Machine Learning
- 23DS610 – Deep Learning
- 23DS620 – Data Mining

**Course Outcomes**

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

CO1	Apply the fuzzy logic operations and the process of fuzzification and defuzzification with a knowledge of fuzzy set theory	Apply
CO2	Apply fuzzy inference systems models, in practical scenarios for decision-making	Apply
CO3	Demonstrate various encoding techniques and reproduction methods in genetic algorithms	Apply
CO4	Use Genetic algorithm in solving optimization problem with different operators	Analyze
CO5	Select appropriate meta heuristics techniques, such as Simulated annealing, Tabu search, PSO, ACO and Bee colony optimization for specific applications	Analyze
CO6	Design and implement hybrid systems with neuro fuzzy modeling	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	M	M	M	L	L	L	L	L	-	L	S	S
CO2	S	S	M	S	M	L	L	L	L	L	-	L	S	S
CO3	S	S	M	M	M	L	L	L	L	L	-	L	S	S
CO4	S	S	S	S	M	L	L	L	L	L	-	L	S	S
CO5	S	S	S	S	M	L	L	L	L	L	-	L	S	S
CO6	S	S	M	M	M	L	L	L	L	L	-	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	5	10	20							25															4		10			
CO2	5	10	20							50															2	5	10			
CO3		10	20							25															4	5	10			
CO4													5	10		20						50		2				10		
CO5													5	10		20						25		4	5		10			
CO6													10	20							25			4	5	10				
Total (in %)	100						100						100						100						100					

## Syllabus

**INTRODUCTION:** Overview of soft computing and historical development, Components of soft computing, Application of soft computing techniques in different areas

**FUZZY SET THEORY:** Basic Definitions and Terminologies, Fuzzy sets and relations, Membership functions, Fuzzification and Defuzzification techniques, Logic and Fuzzy systems, Applications of fuzzy inference systems in decision making

**GENETIC ALGORITHM:** Basic Concepts, Encoding: Binary, Permutation, Tree, Value , Fitness Function, Reproduction: Roulette Wheel, Boltzmann, Tournament, Rank, steady-state, Elitism, Operators : Selection, Crossover(Single point, Two point, Multi point, Uniform, Matrix, Partially Matched, Order and Cycle), Mutation (Flip, Swap, Inverse) – Applications to solve optimization problems

**METAHEURISTIC TECHNIQUES:** Simulated Annealing, Tabusearch, Particle swarm optimization (PSO), Ant colony optimization (ACO), Artificial Bee Colony Optimization - Applications

**HYBRID SYSTEMS:** ANFIS, Hybrid learning, Learning methods, Coactive Neuro Fuzzy modeling, Case studies

## Reference Books & web resources

1. Ross Timothy J., “Fuzzy Logic with Engineering Applications”, John Wiley and Sons, 2010 (Chapters 2,3,4,5)
2. Rajasekaran S., Vijayalakshmi Pai G. A., Neural Networks, Fuzzy Logic and Genetic Algorithms, Prentice Hall, 2006 (Chapters 8,9)
3. Carlos, David, Gary, “Evolutionary algorithms for solving multi-objective problems”, Springer 2007 (Chapters 10)
4. Jang J. S. R., Sun C. T. and Mizutani E., “Neuro-fuzzy and Soft Computing”, PrenticeHall, 2010. (Chapters 12, 13)
5. David E. Goldberg, “Genetic Algorithms In Search, Optimization And Machine Learning”, Pearson Education, 2002

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>INTRODUCTION</b>	
1.1	Overview of soft computing and historical development	1
1.2	Components of soft computing	1
1.3	Application of soft computing techniques in different areas	1
<b>2</b>	<b>FUZZY SET THEORY</b>	
2.1	Basic Definitions and Terminologies	1
2.2	Fuzzy sets and relations	1
2.3	Membership functions	1
2.4	Fuzzification and Defuzzification techniques	3
2.5	Logic and fuzzy systems	2
2.6	Applications of fuzzy logic in decision-making	1

Module No.	Topic	No. of Periods
<b>3</b>	<b>GENETIC ALGORITHM</b>	
3.1	Basic Concepts	1
3.2	Encoding: Binary, Permutation, Tree, Value , Fitness Function	1
3.3	Reproduction: Roulette Wheel, Boltzmann, Tournament, Rank, steady-state, Elitism	2
3.4	Operators : Selection, Crossover(Single point, Two point, Multi point, Uniform, Matrix, Partially Matched, Order and Cycle), Mutation (Flip, Swap, Inverse)	2
3.5	Applications to solve optimization problems	2
<b>4</b>	<b>META HEURISTIC TECHNIQUES</b>	
4.1	Simulated Annealing, Tabusearch	1
4.2	Particle swarm optimization (PSO)	1
4.3	Ant colony optimization (ACO)	2
4.4	Artificial Bee colony	2
4.5	Applications in clustering and routing problems	2
<b>5</b>	<b>HYBRID SYSTEMS</b>	
5.1	ANFIS – Hybrid Learning , Learning methods	2
5.2	Coactive Neuro Fuzzy modeling	2
5.3	Case studies	2
	<b>Total</b>	<b>34</b>

**Course Designer(s):**

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**23DSPH0 MATHEMATICAL MODELING**

Category L T P Credit  
PEC 3 0 0 3

**Preamble**

The students will be able to learn the basic concepts of modelling and its different kinds, Fit data with different models for univariate and multivariate, mean variance portfolio optimization, Markowitz model and efficient frontier calculation algorithm. Also apply the concepts of fundamental theorem of asset pricing and Cox-Ross- Rubinstein (CRR) model, Brownian motion, martingales, risk neutral measure, Girsanov's theorem, Feynman-Kac formula, and bioinformatics.

**Prerequisite**

- 23DS110 Calculus
- 23DS220 Applied Statistics

**Course Outcomes**

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

CO1	Apply the concept of modelling in Fitting data with polynomials, splines	Apply
CO2	Apply the concept of modelling with regression analysis and time series models for univariate and multivariate	Apply
CO3	Apply the concepts of Mean variance portfolio optimization, Markowitz model and efficient frontier calculation algorithm	Apply
CO4	Apply the concepts of fundamental theorem of asset pricing and Cox-Ross-Rubinstein (CRR) model	Apply
CO5	Apply the concepts of Brownian motion, martingales, risk neutral measure, Girsanov's theorem, Feynman-Kac formula	Apply
CO6	Apply the concepts of bioinformatics and model the real world problem using genetic algorithm and hidden markov models	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	S	M	L	L	L	L	-	-	-	-	S	M
CO2	S	S	S	M	L	M	M	L	L	-	L	L	S	S
CO3	S	M	S	M	M	L	L	-	L	-	L	L	S	M
CO4	S	M	S	M	M	L	M	L	-	-	L	-	S	S
CO5	S	M	S	S	L	L	M	-	L	-	L	-	S	S
CO6	S	S	S	S	M	M	L	-	L	-	L	-	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	4	10	20						30																4	10				
CO2	4	10	20						30																2		14			
CO3	2		30						40																	5	14			
CO4													4	10	20									2		14				
CO5													4	10	20										5	14				
CO6													2		30									2		14				
Total (in %)	100						100						100						100											

## Syllabus

**EMPIRICAL MODELING WITH DATA FITTING:** Introduction to Modeling - Modeling process, Overview of different kinds of model. Error function, least squares method; fitting data with polynomials and splines.

**CAUSAL MODELING AND FORECASTING:** Introduction, Modeling the causal time series, forecasting by regression analysis, predictions by regression. Planning, development and maintenance of linear models, trend analysis, modeling seasonality and trend, trend removal and cyclical analysis, decomposition analysis. Modeling financial time series. Econometrics and time series models. Non seasonal models: ARIMA process for univariate and multivariate.

**PORTFOLIO MODELING AND ANALYSIS:** Portfolios, returns and risk, risk- reward analysis, asset pricing models, mean variance portfolio optimization, Markowitz model and efficient frontier calculation algorithm, Capital Asset Pricing Models (CAPM).

**DISCRETE-TIME FINANCE:** Pricing by arbitrage, risk-neutral probability measures, valuation of contingent claims, and fundamental theorem of asset pricing, Cox- Ross-Rubinstein (CRR) model, pricing and hedging of European and American derivatives as well as fixed-income derivatives in CRR model, general results related to prices of derivatives.

**STOCHASTIC CALCULUS:** Brownian motion, martingales, Itô's formula, Itô integral, risk-neutral measure, SDE; Risk-neutral measure, Girsanov's theorem for change of measure, martingale representation theorems, representation of Brownian martingales, Feynman-Kac formula.

**MODELING WITH BIOINFORMATICS:** Introduction, Biological data- types, mode of collection, documentation and submission. Sequence alignment- Definition, significance, dot matrix method, dynamic programming- Global and local alignment tools, scoring matrices and gap penalties. Multiple sequence alignment: Iterative methods. Genetic algorithm, Hidden Markovian models, statistical methods, position specific scoring matrices.

## Reference Books & web resources

1. B. Barnes and G. R. Fulford (2021), "Mathematical Modelling with Case Studies: Using Maple and MATLAB, Third Edition".
2. Edward A. Bender (2022), "An Introduction to Mathematical Modeling: A Course in Mechanics"
3. Christos H. Skiadas and Ioannis Dimotikalis (2021), "Mathematical Modelling and Applications in Elements of Finance".
4. Vladimir Mityushev and Wojciech Nawalaniec (2022), "Introduction to Mathematical Modeling and Computer Simulations".
5. Cornelis W. Oosterlee and Lech A. Grzelak (2019) "Mathematical Modelling and Computation in Finance: With Exercises and Python and MATLAB Computer Codes"
6. Giordano F. R., Weir M. D., and Fox W. P., "A First Course in Mathematical Modeling". Brooks/Cole, Belmont, 2014.
7. Christoffersen P., "Elements of Financial Risk Management", Academic Press, 2012.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>EMPIRICAL MODELING WITH DATA FITTING</b>	
1.1	Introduction to Modeling - Modeling process	1
1.2	Overview of different kinds of model.	1
1.3	Error function	1
1.4	Least squares method	2

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
1.5	Fitting data with polynomials and splines	2
<b>2</b>	<b>CAUSAL MODELING AND FORECASTING</b>	
2.1	Introduction, Modeling the causal time series	1
2.2	Forecasting by regression analysis, predictions by regression.	1
2.3	Planning, development and maintenance of linear models,	1
2.4	Trend analysis, modeling seasonality and trend, trend removal and cyclical analysis, decomposition analysis	1
2.5	Modeling financial time series.	1
2.6	Econometrics and time series models.	1
2.7	Non seasonal models: ARIMA process for univariate and multivariate	2
<b>3</b>	<b>PORTFOLIO MODELING AND ANALYSIS</b>	
3.1	Portfolios, returns and risk, risk- reward analysis	1
3.2	Asset pricing models, mean variance portfolio optimization	1
3.3	Markowitz model and efficient frontier calculation algorithm	1
3.4	Capital Asset Pricing Models (CAPM)	1
<b>4</b>	<b>DISCRETE-TIME FINANCE</b>	
4.1	Pricing by arbitrage, risk-neutral probability measures	1
4.2	Valuation of contingent claims, and fundamental theorem of asset pricing	1
4.3	Cox- Ross-Rubinstein (CRR) model	1
4.4	Pricing and hedging of European and American derivatives as well as fixed-income derivatives in CRR model, general results related to prices of derivatives	2
<b>5</b>	<b>STOCHASTIC CALCULUS</b>	
5.1	Brownian motion, martingales	1
5.2	Itô's formula, Itô integral, risk-neutral measure, SDE	1
5.3	Risk-neutral measure, Girsanov's theorem for change of measure	2
5.4	Martingale representation theorems, representation of Brownian martingales, Feynman-Kac formula	2
<b>6</b>	<b>MODELING WITH BIOINFORMATICS</b>	
6.1	Introduction, Biological data- types, mode of collection, documentation and submission	1
6.2	Sequence alignment- Definition, significance, dot matrix method	1
6.3	Dynamic programming- Global and local alignment tools, scoring matrices and gap penalties	1
6.4	Multiple sequence alignment: Iterative methods	1

Module No.	Topic	No. of Periods
6.5	Genetic algorithm, Hidden Markovian models, statistical methods, position specific scoring matrices	2
	<b>Total</b>	<b>36</b>

**Course Designer(s):**

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**23DSPJ0****GRAPH ALGORITHMS**

Category L T P Credit

PEC 3 0 0 3

**Preamble**

This course will enable students to learn basic concepts of graphs and time-space complexity. Apply the different types of algorithms of shortest path for graphs. Apply the concepts of physical design algorithms and to learn partitioning of circuits. Learn the Global routing algorithms.

**Prerequisite**

- 23DS230: Graph Theory
- 23DS430: Design and Analysis of Algorithms

**Course Outcomes**

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

CO1	Understand the basic concepts of graphs with algorithm time space complexity	Understand
CO2	Apply the different types of algorithms for shortest path of various kinds of graphs.	Apply
CO3	Apply and model the plane sweep algorithm, Max min cut theorem and The Edmonds – Karp algorithm	Apply
CO4	Design the concepts of Physical algorithms.	Apply
CO5	Apply the concept of Partitioning the circuits.	Apply
CO6	Apply the different types of Global routing algorithms for Integer Programming Problem.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	L	M	M	-	L	L	L	M	M	-	S	M
CO2	S	S	M	M	M	L	-	-	M	M	M	M	M	S
CO3	S	S	S	S	M	-	L	M	M	M	M	L	S	M
CO4	M	M	S	-	M	-	M	L	L	L	M	L	M	S
CO5	M	M	S	M	M	M	L	M	M	-	M	-	M	S
CO6	S	M	S	-	S	M	L	M	M	M	M	M	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	4	10						20																	2	10				
CO2	2		40					20	30																2		20			
CO3	4	10	30						30																2	5	10			
CO4													4	10	20					10	20				2		20			
CO5													2		30					10	20				2		10			
CO6													4	10	20					20	20					5	10			
Total(in %)	100						100						100						100											

## Syllabus

**BASIC CONCEPTS:** Graphs – representations, planar graphs- Euler’s formula, crossing number, doubly connected edge list data structure, Algorithm complexity – time and space.

**SHORTEST PATH ALGORITHMS:** Properties for shortest path - The Bellman-Ford algorithm – Single source shortest path in directed acyclic graphs – Dijkstra’s algorithm – Difference constraints and shortest paths – shortest paths and matrix multiplication – The Floyd Warshall Algorithm – Johnson’s algorithm for sparse graphs.

**NETWORK FLOW:** Flow networks and Flows – Flow networks with multiple sources and sinks, The Ford – Fulkerson Method, Augmenting paths – Max Flow min cut theorem, The Edmonds – Karp algorithm.

**PHYSICAL DESIGN:** Classes of graphs in Physical design – Relationship between Graph Classes – Algorithms for Interval graphs, Permutation graphs & Circle graphs.

**PARTITIONING:** Classification of Partitioning Algorithms – Kernighan-Lin Algorithm – Fiduccia-Mattheyses Algorithm - Goldberg and Burstein Algorithm – Component Replication – Ratio Cut.

**GLOBAL ROUTING:** Classification of Global Routing Algorithm – Lee’s Algorithm, Soukup’s Algorithm – Hadlock’s Algorithm – Maze routing Algorithm – Steiner Tree Algorithms.

## Reference Books & web resources

1. Thomas H. Cormen, Charles E. Leiserson, and Ronald LRivest, — Introduction to Algorithms, Fourth Edition, MIT Press, 2022.
2. Naveed A. Sherwani - Algorithms for VLSI Physical Design Automation Springer Publication, 2013.
3. Ravindra K. Ahuja, Thomas L. Magnanti, and James B. Orlin Network Flows: Theory, Algorithms, and Applications, Prentice Hall, 1993.
4. [https://onlinecourses.nptel.ac.in/noc24\\_cs70/preview](https://onlinecourses.nptel.ac.in/noc24_cs70/preview)
5. <https://neo4j.com/blog/top-13-resources-graph-theory-algorithms/>

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>BASIC CONCEPTS</b>	
1.1	Graphs – representations, planar graphs	1
1.2	Euler’s formula, crossing number	2
1.3	Algorithm complexity – time and space	2
<b>2</b>	<b>SHORTEST PATH ALGORITHMS</b>	
2.1	Properties for shortest path, The Bellman-Ford algorithm	2
2.2	Single source shortest path in directed acyclic graphs, Dijkstra’s algorithm	1
2.3	Difference constraints and shortest paths – shortest paths and matrix multiplication	2

Module No.	Topic	No. of Periods
2.4	The Floyd Warshall Algorithm, Johnson's algorithm for sparse graphs.	2
<b>3</b>	<b>NETWORK FLOW</b>	
3.1	Flow networks and Flows	1
3.2	Flow networks with multiple sources and sinks	1
3.3	The Ford – Fulkerson Method, Augmenting paths	2
3.4	Max Flow min cut theorem, The Edmonds – Karp algorithm.	2
<b>4</b>	<b>PHYSICAL DESIGN</b>	
4.1	Classes of graphs in Physical design	2
4.2	Relationship between Graph Classes	1
4.3	Algorithms for Interval graphs, Permutation graphs & Circle graphs	3
<b>5</b>	<b>PARTITIONING</b>	
5.1	Classification of Partitioning Algorithms	1
5.2	Kernighan-Lin Algorithm	2
5.3	Fiduccia-Mattheyses Algorithm, Goldberg and Burstein Algorithm.	2
5.4	Component Replication, Ratio Cut.	1
<b>6</b>	<b>GLOBAL ROUTING</b>	
6.1	Classification of Global Routing Algorithm	1
6.2	Lee's Algorithm, Soukup's Algorithm	2
6.3	Hadlock's Algorithm, Maze routing Algorithm	2
6.4	Steiner Tree Algorithms	1
	<b>Total</b>	<b>36</b>

**Course Designer(s):**

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**23DSPK0****EXPLAINABLE ARTIFICIAL  
INTELLIGENCE**Category L T P Credit  
PEC 3 0 0 3**Preamble**

This course provides an overview of Explainable AI (XAI) techniques, focusing on the importance of transparency and interpretability in machine learning models. It introduces interpretable models, model-agnostic methods, and local and global interpretability techniques, with real-world applications.

**Prerequisite**

- 23DS540: Machine Learning
- 23DS610: Deep Learning

**Course Outcomes**

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

CO1	Explain the importance of interpretability in machine learning models and understand the trade-offs with model accuracy.	Apply
CO2	Describe interpretable models such as linear models and decision trees to provide transparent predictions.	Apply
CO3	Implement model-agnostic methods to interpret black-box models using techniques such as Permutation Feature Importance and PDP.	Apply
CO4	Utilize local interpretability techniques, including LIME and SHAP, to provide explanations for individual predictions.	Apply
CO5	Differentiate between local and global interpretability methods and apply them to real-world scenarios.	Analyze
CO6	Assess ethical implications in AI systems, focusing on bias detection and fairness in interpretable models.	Analyze

**Mapping with Programme Outcomes**

Cos	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	L	M	M	L	L	L	L	L	L	L	S	M
CO2	S	S	M	M	L	L	L	L	L	L	L	L	S	M
CO3	S	S	M	M	M	L	L	L	L	L	L	L	S	S
CO4	S	S	M	S	M	L	L	L	L	L	L	L	S	S
CO5	M	S	S	S	L	M	L	L	L	L	L	L	S	M
CO6	S	M	S	M	M	M	M	S	L	L	L	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	10	10	10							30															4		10			
CO2	10	10	20							30															4		10			
CO3		10	20							40															4	5	10			
CO4													10	10	10										4	5	10			
CO5													10	10	10	10					30	20			2	5		10		
CO6														10	10	10					30	20			2	5		10		
Total (in %)	100						100						100						100											

## Syllabus

**INTRODUCTION TO INTERPRETABLE MACHINE LEARNING-** Definition of interpretability, Why interpretability matters in machine learning, Trade-offs between accuracy and interpretability, Case Study: Real-world impacts of interpretability in sensitive domains (e.g., healthcare, finance)

**INTERPRETABLE MODELS-** Linear models and Generalized Linear Models, Decision trees and rule-based models, k-Nearest Neighbors (k-NN), Naive Bayes classifiers, Case Study: Using decision trees to explain credit scoring models

**MODEL-AGNOSTIC INTERPRETABILITY METHODS-** Permutation Feature Importance, Partial Dependence Plots (PDP), Accumulated Local Effects (ALE), Individual Conditional Expectation (ICE) plots, Case Study: Applying permutation feature importance in a fraud detection model

**LOCAL INTERPRETABILITY TECHNIQUES-** Local Surrogate Models (LIME), Shapley Values and SHAP explanations, Counterfactual Explanations, Anchors for decision rules, Case Study: Counterfactual explanations for loan approval decisions in banking systems

**GLOBAL VS. LOCAL INTERPRETABILITY-** Differences between global and local methods, Global surrogate models, Visualizing machine learning models, Feature importance and feature interactions, Case Study: Global vs. local interpretability for image classification in healthcare applications

**ETHICAL IMPLICATIONS AND CHALLENGES-** Interpretability in healthcare, finance, and legal systems, Bias detection and fairness in AI, Ethical challenges in deploying interpretable models, Future directions in Explainable AI research, Case Study: Ethical implications of explainable AI in healthcare diagnosis models

## Reference Books & web resources

1. Molnar, Christoph. "Interpretable Machine Learning: A Guide for Making Black Box Models Explainable", 2022. (Chapters 1 to 9).
2. Michael Munn and David Pitman, "Explainable AI for Practitioners", O'Reilly Media, 2022.
3. Thampi A. "Interpretable AI: Building Explainable Machine Learning Systems", Leanpub, 2021.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>INTRODUCTION TO INTERPRETABLE MACHINE LEARNING</b>	
1.1	Definition of interpretability	1
1.2	Why interpretability matters in machine learning	1
1.3	Trade-offs between accuracy and interpretability	1
1.4	Case Study: Real-world impacts of interpretability in sensitive domains (e.g., healthcare, finance)	1
<b>2</b>	<b>INTERPRETABLE MODELS</b>	
2.1	Linear models and Generalized Linear Models	2

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
2.2	Decision trees and rule-based models	1
2.3	k-Nearest Neighbors (k-NN)	1
2.4	Naive Bayes classifiers	1
2.5	Case Study: Using decision trees to explain credit scoring models	1
<b>3</b>	<b>Model-Agnostic Interpretability Methods</b>	
3.1	Permutation Feature Importance	2
3.2	Partial Dependence Plots (PDP)	2
3.3	Accumulated Local Effects (ALE)	1
3.4	Individual Conditional Expectation (ICE) plots	1
3.5	Case Study: Applying permutation feature importance in a fraud detection model	1
<b>4</b>	<b>Local Interpretability Techniques</b>	
4.1	Local Surrogate Models (LIME)	2
4.2	Shapley Values and SHAP explanations,	1
4.3	Counterfactual Explanations	1
4.4	Anchors for decision rules	1
4.5	Case Study: Counterfactual explanations for loan approval decisions in banking systems	1
<b>5</b>	<b>Global vs. Local Interpretability</b>	
5.1	Differences between global and local methods	2
5.2	Global surrogate models	1
5.3	Visualizing machine learning models	1
5.4	Feature importance and feature interactions	1
5.5	Case Study: Global vs. local interpretability for image classification in healthcare applications	1
<b>6</b>	<b>Ethical Implications and Challenges</b>	
6.1	Interpretability in healthcare, finance, and legal systems	2
6.2	Bias detection and fairness in AI	1
6.3	Ethical challenges in deploying interpretable models	1
6.4	Future directions in Explainable AI research	1
6.5	Case Study: Ethical implications of explainable AI in healthcare diagnosis models	1
	<b>Total</b>	<b>35</b>

**Course Designer:**

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**23DSPL0****GAME THEORY**

Category L T P Credit

PEC 3 0 0 3

**Preamble**

It equips students with foundational and advanced concepts in strategic decision-making. Through topics like Nash equilibrium, cooperative games, and mechanism design, students will gain practical skills to apply game theory in data-driven fields such as economics, business, and algorithmic decision-making.

**Prerequisite**

- 23DS210: Theory of Probability
- 23DS410: Linear Algebra

**Course Outcomes**

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

CO1	Explain foundational concepts of game theory, including Nash equilibrium and strategic form games.	Understand
CO2	Analyze extensive form games and their applications in decision-making.	Analyze
CO3	Apply cooperative game theory concepts, such as the Shapley value and core.	Apply
CO4	Apply principles from mechanism design theory to design effective mechanisms.	Apply
CO5	Apply models of bargaining and negotiation and how they can be applied to models of competition.	Apply
CO6	Analyze the societal implications of game-theoretic models in real-world scenarios.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	S	M	M	-	-	L	-	-	-	-	S	M
CO2	S	S	M	S	M	L	-	L	-	-	-	-	S	M
CO3	S	S	M	S	-	-	L	-	-	-	-	-	S	M
CO4	M	M	S	M	S	-	-	-	-	-	L	-	S	M
CO5	M	M	S	S	S	-	-	-	-	-	-	L	S	M
CO6	S	S	S	M	-	S	L	-	-	M	-	-	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	10	20						25																	2	10				
CO2									25				10				20						50		2					15
CO3													5	10	20								25		2	5	10			
CO4													5	10	20								25		2	5	10			
CO5	5	10	20																						2	5	10			
CO6	5	10		20						50																5		15		
Total (in %)	100						100						100						100						100					

**Syllabus**

**INTRODUCTION TO GAME THEORY:** Definition of Game Theory -Rational Choice and Utility Functions - Strategic vs. Extensive Form Games - Examples: Cournot and Bertrand competition models

**STRATEGIC FORM GAMES:** Pure and Mixed Strategy Nash Equilibria - Examples from economics, business, environment, military, market - Best Response Functions and Dominated Actions - Existence and Computation of Nash Equilibria - Zero-Sum Games and Minimax Theorem - Illustrations: Electoral competition, Vickrey auctions, Cournot competition

**EXTENSIVE FORM GAMES:** Games with Perfect Information -Backward Induction -Nash and Subgame Perfect Equilibrium - Repeated Games (Introduction) - Illustrations: Sequential bargaining, monopolistic industry entry

**COOPERATIVE GAMES:** Transferable Utility Games - The Core and Shapley Value -Nash Bargaining Solution

**MECHANISM DESIGN AND AUCTIONS:** Introduction to Mechanism Design -Social Choice Functions and Incentive Compatibility - Gibbard-Satterthwaite Theorem -Vickrey-Clarke-Groves (VCG) Mechanism - Optimal Mechanisms and Myerson Auction

**ADVANCED TOPICS:** Bayesian Games and Incomplete Information - Evolutionary Stable Strategies (ESS) - Applications to Population Games

**Reference Books & web resources**

- Noam Nisan, Tim Roughgarden, Eva Tardos, Vijay V. Vazirani, (2011). "Algorithmic Game Theory". Cambridge University Press.
- Osborne, M. J. (2004). "An Introduction to Game Theory". Oxford University Press google scholar, 2, 672-713. (Chapter: 2, 4, 9)
- Anna R. Karlin and Yuval Peres, "Game Theory, Alive," American Mathematical Society, Apr 27, 2017, ISBN-13: 978-1470419820.
- Martin J. Osborne and Ariel Rubinstein, "A Course in Game Theory", MIT Press, 2016 (New Edition)
- Siddharth Barman and Y. Narahari, "Game Theory Lecture Notes" Available online at <https://lcm.csa.iisc.ernet.in/gametheory>
- Kevin Leyton-Brown and Yoav Shoham, "Essentials of Game Theory: A Concise, Multidisciplinary Introduction", Morgan & Claypool Publishers, 2016
- E. N. Barron, "Game Theory: An Introduction" (2nd Edition), Wiley, 2017

**Course Contents and Lecture Schedule**

Module No.	Topic	No. of Periods
<b>1</b>	<b>INTRODUCTION TO GAME THEORY</b>	
1.1	Definition of Game Theory	1
1.2	Rational Choice and Utility Functions	1
1.3	Strategic vs. Extensive Form Games	1
1.4	Examples: Cournot and Bertrand competition models	2
<b>2</b>	<b>STRATEGIC FORM GAMES</b>	
2.1	Pure and Mixed Strategy Nash Equilibria	2

Module No.	Topic	No. of Periods
2.2	Examples from economics, business, environment, military, market	1
2.3	Best Response Functions and Dominated Actions	1
2.4	Existence and Computation of Nash Equilibria	2
2.5	Zero-Sum Games and Minimax Theorem	1
2.6	Illustrations: Electoral competition, Vickrey auctions, Cournot competition	2
<b>3</b>	<b>EXTENSIVE FORM GAMES</b>	
3.1	Games with Perfect Information	1
3.2	Backward Induction	1
3.3	Nash and Subgame Perfect Equilibrium	2
3.4	Repeated Games (Introduction)	1
3.5	Illustrations: Sequential bargaining, monopolistic industry entry	2
<b>4</b>	<b>COOPERATIVE GAMES</b>	
4.1	Transferable Utility Games	1
4.2	The Core and Shapley Value	2
4.3	Nash Bargaining Solution	1
<b>5</b>	<b>MECHANISM DESIGN AND AUCTIONS</b>	
5.1	Introduction to Mechanism Design	1
5.2	Social Choice Functions and Incentive Compatibility	1
5.3	Gibbard-Satterthwaite Theorem	1
5.4	Vickrey-Clarke-Groves (VCG) Mechanism	1
5.5	Optimal Mechanisms and Myerson Auction	2
<b>6</b>	<b>ADVANCED TOPICS</b>	
6.1	Bayesian Games and Incomplete Information	1
6.2	Evolutionary Stable Strategies (ESS)	1
6.3	Applications to Population Games	1
	<b>Total</b>	<b>34</b>

**Course Designer(s):**

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**23DSPM0 SOCIAL MEDIA ANALYTICS**

Category L T P Credit  
PEC 3 0 0 3

**Preamble**

This course aims to provide students with the knowledge of key concepts in social media analytics, focusing on the representation of knowledge using ontology and the concept of the Semantic Web and explore the practical and theoretical aspects of analyzing social network data, performing extraction, reasoning, and applying it to real-world optimization, prediction, and evaluation problems.

**Prerequisite**

- 23DS230 - Graph Theory
- 23DS420 - Predictive Analytics

**Course Outcomes**

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

CO1	Understand and explain the basic concepts of the Semantic Web.	Understand
CO2	Describe key concepts in network analysis	Understand
CO3	Apply data extraction techniques to the social network data	Apply
CO4	Analyse and examine with aggregated social network data using ontologies.	Analyze
CO5	Compare and contrast various methods in knowledge representation for social networks.	Apply
CO6	Identify the application area of optimization, prediction and evaluation in real world scenario.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	L	L	M	L	L	M	-	L	S	M	L	M
CO2	S	S	M	L	M	L	M	M	-	L	S	L	M	L
CO3	M	S	S	M	L	M	-	-	-	L	S	M	L	M
CO4	M	M	S	S	M	S	L	-	-	L		S	L	M
CO5	L	M	M	M	S	M	L	M	-	L	L	S	S	M
CO6	L	M	M	M	S	M	-	-	-		S	S	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	10	20						20																	5	10				
CO2	10	20						30																	5	10				
CO3	5	5	30						50																			15		
CO4													5	5	20							40						20		
CO5													10	20							20						15			
CO6													10	10	20						40							20		
Total (in %)	100						100						100						100											

## Syllabus

**INTRODUCTION TO THE SEMANTIC WEB** : Limitations of the Current Web - Address issues: lack of structured data, semantic ambiguity, and human dependency for data interpretation - Development of the Semantic Web- technologies RDF and OWL - emergence of the Social Web- Explore transformation and content sharing of Facebook.

**SOCIAL NETWORK ANALYSIS** : Development of Social Network Analysis- Historical perspective of SNA and its applications- sociology and marketing -Key Concepts and Measures in Network Analysis- metrics of centrality and closeness centrality- concept of weak ties - Electronic Sources for Network Analysis.

**KNOWLEDGE REPRESENTATION ON THE SEMANTIC WEB** : Ontology and Role of Semantic Web - frameworks of knowledge and retrieval information - Ontology-Based Knowledge Representation- improve machine understanding through ontologies structure data-Ontology Languages for the Semantic Web: RDF, OWL, and SPARQL.

**MODELING AND AGGREGATING SOCIAL NETWORK DATA:** State-of-the-Art in Network Data Representation - contemporary methods represent social relationships and interactions- Ontological Representation of Social Individuals and Relationships- social entities and attributes modeled - aggregating and Reasoning with Social Network Data- Techniques for synthesizing data.

**SOCIAL-SEMANTIC APPLICATIONS AND DATA EXTRACTION:** Generic Architecture for Social-Semantic Applications- Understanding frameworks - integrate social and semantic data- Tools and Technologies such as Apache Jena, Neo4j, and D3.js - Social Network Extraction- web scraping and API usage- Optimization, Prediction, and Evaluation- Explore machine learning methods for predictive analytics - evaluation of social media campaigns.

## Reference Books & web resources

1. John G. Breslin, Alexandre Passant, Stefan Decker, "The Social Semantic Web", Springer, 2009.(Chapters 1-8)
2. Yong Wang, Wei Fan, Zhiyong Lu, "Social Semantic Web: Technology and Applications", 2010.
3. Simon Haykin, "Digital Communications", John Wiley & Sons Pvt. Ltd., 2001.
4. Dean Allemang & James Hendler, "Semantic Web Foundations: Semantic Web for the Working Ontologist", 2011.
5. John G. Proakis, "Digital Communications" McGraw Hill International Edition, Fourth Edition, 2001.
6. Bernard Sklar, "Digital Communications: Fundamentals and Applications", 2<sup>nd</sup> Edition, Prentice Hall, 2001.
7. John R Barry, Edward Lee and David G, Messerschmitt, "Digital Communication", 3<sup>rd</sup> Edition, Springer, 2003.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>INTRODUCTION TO THE SEMANTIC WEB</b>	
1.1	Limitations of the Current Web	2
1.2	Development of the Semantic Web	2

Module No.	Topic	No. of Periods
1.3	Emergence of the Social Web	2
<b>2</b>	<b>SOCIAL NETWORK ANALYSIS</b>	
2.1	Development of Social Network Analysis	2
2.2	Key Concepts and Measures in Network Analysis	3
2.3	Electronic Sources for Network Analysis	2
<b>3</b>	<b>KNOWLEDGE REPRESENTATION ON THE SEMANTIC WEB</b>	
3.1	Ontology and Their Role in the Semantic Web	2
3.2	Ontology-Based Knowledge Representation	2
3.3	Ontology Languages for the Semantic Web (RDF and OWL)	2
<b>4</b>	<b>MODELLING AND AGGREGATING SOCIAL NETWORK DATA</b>	
4.1	State-of-the-Art in Network Data Representation	2
4.2	Ontological Representation of Social Individuals and Relationships	3
4.3	Aggregating Social Network Data	2
4.4	Reasoning with Social Network Data	2
<b>5</b>	<b>SOCIAL-SEMANTIC APPLICATIONS AND DATA EXTRACTION</b>	
5.1	Generic Architecture for Social-Semantic Applications	2
5.2	Tools and Technologies for Data Processing (Apache Jena, Neo4j, D3.js)	3
5.3	Social Network Extraction Techniques (Web Scraping and APIs)	2
	<b>Total</b>	<b>34</b>

**Course Designer(s):**

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**23DSPN0****CLOUD COMPUTING**

Category L T P Credit

PEC 3 0 0 3

**Preamble**

This course offers a comprehensive exploration of virtualization and cloud computing concepts, focusing on the technologies and architectures that enable modern cloud services. Also focus on virtualization and hypervisors, delve into cloud architecture fundamentals, service models, and storage solutions, providing insights into the intricacies of cloud deployments and security considerations will be addressed in detail.

**Prerequisite**

- 23DS550 - Computer Networks

**Course Outcomes**

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

CO1	Understand the foundational concepts of virtualization, hypervisor types and roles	Understand
CO2	Implement the cloud architecture fundamentals, including service models and deployment strategies.	Apply
CO3	Integrate various cloud service models, including SaaS, PaaS, and IaaS, and their real-world applications.	Apply
CO4	Implement the security challenges	Apply
CO5	Utilize and implement deployment tools and techniques	Apply
CO6	Explore and analyse advanced cloud concepts and conduct case studies.	Analyse

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	S	M	L	L	M	-	M	-	-	L	L	M
CO2	S	S	S	M	M	M	M	-	S	-	-	L	M	L
CO3	M	S	S	S	S	M	M	-	M	-	-	M	L	M
CO4	M	M	S	S	M	M	S	-	M	-	-	S	L	L
CO5	S	M	S	S	S	S	M	-	S	-	-	M	S	S
CO6	M	S	S	M	S	M	M	-	S	-	-	S	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	10	10																							10	10				
CO2	5	10	30						50																		10			
CO3	5	10	20							50																5	10			
CO4												10			10										5	10				
CO5														10	20					50						10				
CO6														10	20	20					50					10	20			
Total (in %)	100						100						100						100											

## Syllabus

**VIRTUALIZATION AND Hypervisors** : Fundamental concepts of virtualization - significance of modern computing - types of virtualization - overview of hypervisors-architecture - functionalities of Type 1 (bare-metal) and Type 2 (hosted) hypervisors - various hypervisors (Xen, KVM, VMware, VirtualBox, and Hyper-V) - role of hypervisors - use cases- performance characteristics - management tools.

**CLOUD ARCHITECTURE FUNDAMENTALS** : Focusing on cloud architecture - core characteristics - service models of Infrastructure as a Service (IaaS), Platform as a Service (PaaS), and Software as a Service (SaaS) - deployment models (public, private, hybrid, and community clouds) - essential cloud concepts and terminologies - load balancing – scalability – elasticity - deployment strategies - monitoring cloud environments - Service Level Agreements (SLAs) - billing practices in cloud computing.

**CLOUD SERVICE MODELS AND STORAGE SOLUTIONS** : SaaS segment - Open SaaS approach- PaaS evolution, benefits, and disadvantages - IaaS component - performance strategies - system redundancy- cloud-based NAS devices - Identity as a Service (IDaaS)- Database as a Service (DBaaS) - Monitoring as a Service (MaaS) - different cloud storage providers - cloud file systems - techniques MapReduce - Case studies : Amazon S3 and Walrus.

**SECURITY AND DEPLOYMENT TOOLS IN CLOUD COMPUTING** : critical aspects of security - identity and access management - data loss prevention - web security and security assessments - intrusion management, and encryption - business continuity and disaster recovery (BCDR) implementations - network security challenges in cloud environments - cloud deployment tools of Eucalyptus, Nimbus, OpenStack, CloudStack, and OpenNebula.

**ADVANCED CLOUD CONCEPTS AND CASE STUDIES** : Advanced cloud concepts - cloud operating systems - federated clouds - data center architectures tailored for cloud computing - principles of virtualization platforms - virtual machine migration, and load balancing techniques - scalability, performance, and quality of service (QoS) in cloud computing environments - alongside security and privacy issues - simulation tools like CloudSim.

## Reference Books & Web Resources

1. Matthew Portnoy, "Virtualization Essentials", 2016. (Chapter : 1,2,3,4)
2. Arshdeep Bahga and Vijay Madisetti, "Cloud Computing: A Hands-On Approach", 2014.
3. Thomas Erl, "Cloud Computing: Concepts, Technology & Architecture", 2013(Chapter 1,2,3)
4. Michael J. Kavis," Architecting the Cloud: Design Decisions for Cloud Computing Service Models (SaaS, PaaS, and IaaS)", 2014.(Chapter 2,3,5,7,9)
5. [https://onlinecourses.nptel.ac.in/noc21\\_cs14/preview](https://onlinecourses.nptel.ac.in/noc21_cs14/preview), By Prof. Soumya Kanti Ghosh, IIT Kharagpur, NPTEL.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>VIRTUALIZATION AND HYPERVISORS</b>	
1.1	Introduction to Virtualization	1
1.2	Types of Virtualization	1
1.3	Overview of Hypervisors	1
1.4	In-depth Analysis of Hypervisors	2
1.5	Hypervisors in Cloud Computing	1
<b>2</b>	<b>CLOUD ARCHITECTURE FUNDAMENTALS</b>	
2.1	Introduction to Cloud Architecture	1
2.2	Service Models: IaaS, PaaS, SaaS	2
2.3	Deployment Models: Public, Private, Hybrid, and Community Clouds	2
2.4	Scalability, Elasticity, Load Balancing, and Monitoring in Cloud Environments	1
2.5	Software-Defined Networking (SDN) and Network Function Virtualization (NFV)	1
<b>3</b>	<b>CLOUD SERVICE MODELS AND STORAGE SOLUTIONS</b>	
3.1	SaaS: Multitenancy, Open SaaS Approach	2
3.2	PaaS: Evolution, Benefits, and Limitations	2
3.3	IaaS: Performance Strategies, System Redundancy, Cloud-based NAS	2
3.4	Identity as a Service (IDaaS), Database as a Service (DBaaS), Monitoring as a Service (MaaS)	2
3.5	Cloud File Systems and MapReduce	1
<b>4</b>	<b>SECURITY AND DEPLOYMENT TOOLS IN CLOUD COMPUTING</b>	
4.1	Security in Cloud: Identity Management, Data Loss Prevention	1
4.2	Encryption Techniques and Intrusion Management in Cloud Environments	1
4.3	Business Continuity and Disaster Recovery (BCDR) in Cloud Computing	1
4.4	Cloud Deployment Tools	1
<b>5</b>	<b>ADVANCED CLOUD CONCEPTS AND CASE STUDIES</b>	
5.1	Cloud Operating Systems, Federated Clouds	1
5.2	Data Centre Architectures for Cloud Computing	1
5.3	Virtual Machine Migration and Load Balancing Techniques	1

Module No.	Topic	No. of Periods
5.4	Scalability, Performance, and QoS in Cloud Computing	1
5.5	Case Studies: Amazon Web Services (AWS), Microsoft Azure, IBM Cloud	1
<b>6</b>	<b>SIMULATION TOOLS AND PRESENTATIONS</b>	
6.1	Introduction to CloudSim: Features and Use Cases	1
6.2	Practical Simulation with CloudSim	1
6.3	Group Presentations on Case Studies	1
	<b>Total</b>	<b>34</b>

**Course Designer(s):**

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**23DSPP0****INTERNET OF THINGS AND  
DATA ANALYTICS**

Category	L	T	P	Credit
PEC	2	1	0	3

**Preamble**

This course provides an overview of the Internet of Things (IoT), covering key concepts, architectures, and applications. Students will explore data collection methods, analytics techniques, and the challenges of security and interoperability, preparing them to tackle real-world IoT problems.

**Prerequisite**

- 23DS130: Digital Electronics and Computer Organization
- 23DS470: Predictive Analytics Lab
- 23DS550: Computer Networks
- 23DS680: Big Database Systems Lab

**Course Outcomes**

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

CO1	Describe how IoT systems work and their real-world uses.	Understand
CO2	Analyze IoT data to get useful insights using different analytics methods.	Analyze
CO3	Apply NoSQL databases to store and manage IoT data.	Apply
CO4	Examine IoT solutions by evaluating sensor and device integration	Analyze
CO5	Use machine learning to analyze IoT data for predicting trends and spotting issues.	Analyze
CO6	Identify and solve security and privacy problems in IoT systems.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	S	M	S	-	-	-	-	M	-	L	S	M
CO2	S	S	L	S	M	-	-	-	-	-	-	-	S	M
CO3	S	M	S	M	S	-	L	-	-	-	-	-	S	M
CO4	S	S	S	S	S	S	M	L	S	M	L	M	S	M
CO5	S	S	S	S	M	-	-	-	-	-	-	L	S	M
CO6	S	M	S	M	L	S	L	S	-	M	-	L	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	10	20						25																	2	10				
CO2													10			20									2			20		
CO3	5	10	20						25																		10			
CO4	5	10		20						50															2			20		
CO5													10			20							50		2			20		
CO6													5	10	20								25		2		10			
Total (in %)	100						100						100						100						100					

## Syllabus

**INTRODUCTION TO IOT (INTERNET OF THINGS):** Definition and key concepts - IoT ecosystem - Applications - IoT Architecture (Layers of IoT, IoT Communication protocols) - IoT Challenges (Security and data privacy, scalability, interoperability).

**IOT HARDWARE AND DATA COLLECTION:** IoT Devices and Sensors (Sensors, actuators and microcontrollers) - Data Collection Methods (Real-time data acquisition and streaming) - CaseStudy (smart agriculture and healthcare)

**IOT DATA STORAGE AND MANAGEMENT:** Introduction to NoSQL Databases (MongoDB, Cassandra, HBase) and their relevance for IoT - Data models for IoT (Key-Value, Document, Column-Family, Graph) - Storing and querying IoT data in NoSQL databases - Case Study (Using MongoDB for storing sensor data in a smart city or smart agriculture application)

**IOT DATA ANALYTICS BASICS:** Introduction to IoT Data Analytics and its important - Descriptive (Aggregation and visualization of IoT data) and Predictive Analytics (Predicting sensor failures) - Tools for IoT Analytics (Google BigQuery, Power BI) - Case Study (Predictive maintenance in industrial IoT using sensor data) - Security in IoT Data Analytics (security threats and anomalies).

**EDGE AND FOG COMPUTING IN IOT:** Edge vs. cloud-based data collection - Edge analytics vs.cloud analytics - Bringing analytics closer to IoT devices - Case Study (Smart grids and autonomous vehicles as real-world examples of edge/fog computing applications)

**ADVANCED MACHINE LEARNING FOR IOT DATA:** Introduction to time-series forecasting for IoT sensor data. - Applying ML algorithms (ARIMA, LSTM) to IoT datasets - Real-time data processing and prediction for IoT applications (anomaly detection in IoT networks)

## Reference Books & web resources

1. Amit Kumar Tyagi (2022). "Internet of Things: Theory & Practice". BPB Publications (Chapters: 1, 3, 11 & 14)
2. K.N. Raja Rao (2018). "Internet of Things: Concepts and Applications (An Indian Adaptation)". Wiley (Chapter: 1, 2, 3, 4, 5 & 6)
3. Misra, S., Mukherjee, A., & Roy, A. (2021). "Introduction to IoT". Cambridge University Press. (Chapter: 5, 10, 11, 12, 13, 14, 17, 16)
4. Bahga, A. (2014). "Internet of Things: A Hands-On Approach".
5. Waher, P. (2015). "Learning internet of things". Packt publishing.
6. Pethuru Raj & Anupama C. Raman. "The Internet of Things: Enabling Technologies, Platforms, and Use Cases". CRC Press

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>INTRODUCTION TO IOT</b>	
1.1	Definition and Key Concepts	1
1.2	IoT Ecosystem and Applications	1
1.3	IoT Architecture (Layers of IoT, IoT Communication Protocols)	2
1.4	IoT Challenges (Security and data privacy, scalability, interoperability)	1

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
<b>2</b>	<b>IOT HARDWARE AND DATA COLLECTION</b>	
2.1	IoT Devices and Sensors (Sensors, Actuators, Microcontrollers)	2
2.2	Data Collection Methods (Real-time data acquisition and streaming)	2
2.3	Case Study (Smart agriculture and healthcare)	2
<b>3</b>	<b>IOT DATA STORAGE AND MANAGEMENT</b>	
3.1	Introduction to NoSQL Databases (MongoDB, Cassandra, HBase)	2
3.2	Data Models for IoT (Key-Value, Document, Column-Family, Graph)	2
3.3	Storing and Querying IoT Data in NoSQL Databases	1
3.4	Case Study (Using MongoDB for storing sensor data in a smart city or smart agriculture application)	2
<b>4</b>	<b>IOT DATA ANALYTICS BASICS</b>	
4.1	Introduction to IoT Data Analytics and Its Importance	1
4.2	Descriptive Analytics (Aggregation and Visualization of IoT Data)	1
4.3	Predictive Analytics (Predicting Sensor Failures)	1
4.4	Tools for IoT Analytics (Google BigQuery, Power BI)	2
4.5	Case Study (Predictive maintenance in industrial IoT using sensor data)	1
<b>5</b>	<b>EDGE AND FOG COMPUTING IN IOT</b>	
5.1	Edge vs. Cloud-based Data Collection	1
5.2	Edge Analytics vs. Cloud Analytics	1
5.3	Bringing Analytics Closer to IoT Devices	1
5.4	Case Study (Smart grids and autonomous vehicles)	2
<b>6</b>	<b>ADVANCED MACHINE LEARNING FOR IOT DATA</b>	
6.1	Introduction to Time-Series Forecasting for IoT Sensor Data	1
6.2	Applying ML Algorithms (ARIMA, LSTM) to IoT Datasets	2
6.3	Real-Time Data Processing and Prediction for IoT Applications (Anomaly detection in IoT networks)	2
	<b>Total</b>	<b>34</b>

**Course Designer:**

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**23DSPQ0 COMPUTATIONAL FINANCE**

Category L T P Credit  
PEC 3 0 0 3

**Preamble**

This Computational Finance course teaches current C++ programming and financial modeling. Parallel computing, Monte Carlo simulations, and adjoint differentiation. Students learn how to create high-performance financial simulations for data-driven finance jobs.

**Prerequisite**

- 23DS510 : Numerical Methods
- 23DS520 : Optimization Techniques

**Course Outcomes**

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

CO1	Explain the modern C++ programming concepts for effective financial modeling.	Understand
CO2	Implement parallel computing techniques to enhance the performance of financial algorithms and simulations.	Apply
CO3	Use Monte Carlo simulations for asset pricing and risk assessment in financial applications.	Apply
CO4	Illustrate adjoint differentiation methods for optimizing financial models and calculating sensitivities.	Apply
CO5	Implement the serial and parallel financial simulations using efficient coding practices.	Apply
CO6	Examine the memory management using data structures for algorithmic adjoint differentiation in financial contexts.	Analyze

**Mapping with Programme Outcomes**

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

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	-	S	-	-	-	-	-	M	S	M	S	S
CO2	S	S	M	S	S	-	-	-	-	M	S	S	S	S
CO3	S	S	S	M	S	S	-	-	-	S	S	-	S	M
CO4	-	S	M	S	-	-	S	M	-	-	-	S	S	M
CO5	M	S	M	-	S	S	-	M	-	S	M	-	S	S
CO6	-	M	M	-	M	M	-	S	-	M	-	M	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	10	10						20																	10	10				
CO2	10	10	20						40																	10	10			
CO3		10	30						40																		20			
CO4											10	10	10							20						10				
CO5											10	10	30							40					10	10				
CO6															20						40						10			
Total (in %)	100						100						100						100											

## Syllabus

**MODERN C++ CONCEPTS FOR COMPUTATIONAL FINANCE:** Lambda Expressions - Functional Programming - Move Semantics, Smart Pointers.

**INTRODUCTION TO PARALLEL C++ FOR FINANCIAL APPLICATIONS:** Multi-threading - Thread Management - Data Sharing - Thread Local Storage – Locks – Deadlocks – RAII - Condition Variables - Task Management.

**MONTE CARLO SIMULATIONS FOR ASSET PRICING:** Asset Pricing Models - Monte Carlo Methods - Random Number Generation - Randomness Quality Improvement.

**SERIAL AND PARALLEL IMPLEMENTATION OF FINANCIAL SIMULATIONS:** Simulation Algorithms - Serial Implementation - Parallel Implementation - Skip Ahead Techniques.

**INTRODUCTION TO ADJOINT DIFFERENTIATION FOR FINANCIAL MODELS:** Adjoint Differentiation - Calculation Graphs - Directed Acyclic Graphs - Adjoint Mathematics - Sensitivities.

**ALGORITHMIC ADJOINT DIFFERENTIATION AND EXPRESSION TEMPLATES:** Algorithmic Differentiation - Memory Management - Expression Templates - Templated Code.

## Reference Books

1. Antoine Savine, "Modern Computational Finance: AAD and Parallel Simulations", Hardcover – Illustrated, 7 December 2018.
2. Argimiro Arratia, "Computational Finance: An Introductory Course with R: 1 (Atlantis Studies in Computational Finance and Financial Engineering)", Hardcover – Abridged, 20 May 2014.
3. Omur Ugur, "Introduction to Computational Finance", An: 1 (Series in Quantitative Finance), Hardcover, 29 December 2008, ISBN 978-1848161924, Publisher: Imperial College Press.
4. T. D. M. H. C. N. P. A. P. A., "Quantitative Methods in Derivatives Pricing - An Introduction to Computational Finance": 124 (Wiley Finance), Hardcover, 16 May 2002, ISBN 978-0471394471, Publisher: John Wiley & Sons Inc.
5. C. M. F. L., "Computational Methods for Quantitative Finance: Finite Element Methods for Derivative Pricing (Springer Finance)", Hardcover, March 6, 2015, ISBN 978-3642435324, Publisher: Springer.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>MODERN C++ CONCEPTS</b>	
1.1	Lambda Expressions	2
1.2	Functional Programming	2
1.3	Move Semantics	2
1.4	Smart Pointers	2
2	<b>PARALLEL PROGRAMMING IN C++</b>	
2.1	Introduction to Multithreading	2
2.2	Thread Management	2
2.3	Data Synchronization	1

2.4	Deadlock Prevention	1
2.5	Task Management	2
3	<b>MONTE CARLO METHODS IN FINANCE</b>	
3.1	Asset Pricing Models	2
3.2	Monte Carlo Simulation Techniques	1
3.3	Random Number Generation	1
3.4	Variance Reduction Techniques	1
4	<b>IMPLEMENTATION OF FINANCIAL SIMULATIONS</b>	
4.1	Serial vs. Parallel Algorithms	2
4.2	Monte Carlo Simulation Implementation	2
4.3	Skip-Ahead Techniques	2
5	<b>ADJOINT DIFFERENTIATION</b>	
5.1	Overview of Adjoint Differentiation	2
5.2	Gradient Calculation with AD	1
5.3	Adjoint Methods for Pricing Models	2
6	<b>ALGORITHMIC DIFFERENTIATION TECHNIQUES</b>	
6.1	Efficient AAD Implementation	1
6.2	Expression Templates	1
6.3	Case Studies in Financial Applications	2
	<b>Total</b>	<b>36</b>

**Course Designer(s):**

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**23DSPR0****ENTERPRISE INFORMATION  
SYSTEMS**Category L T P Credit  
PEC 3 0 0 3**Preamble**

This course aims to enable the students to make or influence decisions related to the selection, design and support of Enterprise Information Systems.

**Prerequisite**

- NIL

**Course Outcomes**

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

CO1	Explain the design, construction and management of relational database systems.	Understand
CO2	Describe the strengths and weakness of different approaches to data management.	Understand
CO3	Apply knowledge of processes, tools and techniques involved in information management within an enterprise.	Apply
CO4	Compare and contrast the range of database technologies and methods of database construction.	Apply
CO5	Apply the knowledge of database-driven systems within a business context.	Apply
CO6	Analyze the information and data technologies in the context of organizational needs.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	M	M	M	M	-	-	-	-	-	-	-	L	M
CO2	M	M	M	M	M	-	-	-	-	-	-	-	M	M
CO3	M	S	S	S	M	M	M	L	M	M	-	-	M	S
CO4	M	M	S	M	M	L	-	-	-	M	M	M	M	S
CO5	M	S	S	S	M	L	-	-	-	S	S	S	M	S
CO6	S	S	S	M	M	M	M	M	M	M	M	M	M	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	10	10																							4	6				
CO2	10	10																							4	6				
CO3	10	10	20						50					10											4	6	10			
CO4													10	10	20						40				4	6	10			
CO5													10	10	20						40				4	6	10			
CO6			10	10						50				10								20					10	10		
Total (in %)	100						100						100						100											

## Syllabus

**ORGANIZATIONS, MANAGEMENT, AND THE NETWORKED ENTERPRISE:** the role of information systems in Business today, Perspectives on Information Systems, contemporary approaches to information systems.

**INFORMATION SYSTEMS, ORGANIZATIONS, AND STRATEGY:** organizations and information systems, information systems impact organizations and business firms, using information systems to achieve competitive advantage, using systems for competitive advantage: management issues.

**IT INFRASTRUCTURE AND EMERGING TECHNOLOGIES:** IT infrastructure, infrastructure components, contemporary hardware platform trends, contemporary software platform trends.

**MANAGING KNOWLEDGE:** the knowledge management landscape, enterprise-wide knowledge management systems, knowledge work systems.

**ENHANCING DECISION MAKING:** decision making and information systems, business intelligence in the enterprise, business intelligence constituencies.

**BUILDING INFORMATION SYSTEMS:** Systems as planned organizational change, overview of systems development systems, alternative systems-building approaches, application development for the digital firm.

**MANAGING GLOBAL SYSTEMS:** the growth of international information systems, organizing international information systems.

## Reference Books & web resources

1. Kenneth C. Laudon Jone & P. Laudon, "Management Information Systems", Thirteenth Edition, Pearson Education Limited 2014.
2. Terry Lucey, "Management Information System", Ninth Edition, 2005, Thompson.
3. Effy Oz, "Management Information System", Fourth International Student Edition, Thomson, 6th Edition, 2008.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>ORGANIZATIONS, MANAGEMENT, AND THE NETWORKED ENTERPRISE</b>	
1.1	The role of information systems in Business today	1
1.2	Perspectives on Information Systems, contemporary approaches to information systems.	2
2	<b>INFORMATION SYSTEMS, ORGANIZATIONS, AND STRATEGY</b>	
2.1	Organizations and information systems,	1
2.2	Information systems impact organizations and business firms,	2
2.3	Using information systems to achieve competitive advantage,	1
2.4	Using systems for competitive advantage: management issues.	1
3	<b>IT INFRASTRUCTURE AND EMERGING TECHNOLOGIES</b>	
3.1	IT infrastructure, infrastructure components,	1
3.2	Contemporary hardware platform trends & contemporary software platform trends.	1

3.3	Managing Knowledge	2
3.4	The knowledge management landscape,	1
3.5	Enterprise-wide knowledge management systems,	2
3.6	Knowledge work systems.	2
4	<b>ENHANCING DECISION MAKING</b>	
4.1	Decision making and information systems,	2
4.2	Business intelligence in the enterprise,	2
4.3	Business intelligence constituencies.	2
5	<b>BUILDING INFORMATION SYSTEMS</b>	
5.1	Systems as planned organizational change,	2
5.2	Overview of systems development systems,	2
5.3	Alternative systems-building approaches,	2
5.4	Application development for the digital firm.	2
6	<b>MANAGING GLOBAL SYSTEMS</b>	
6.1	The growth of international information systems,	2
6.2	Organizing international information systems.	1
	<b>Total</b>	<b>34</b>

**Course Designer(s):**

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<b>23DSPS0</b>	<b>RANDOMIZED ALGORITHMS</b>	Category	L	T	P	Credit
		PEC	3	0	0	3

**Preamble**

This course aims to introduce students to randomized algorithms, focusing on their design, classification, and application in optimization problems. Students will develop and analyze efficient algorithms utilizing randomness to solve real-world challenges.

**Prerequisite**

- 23DS430 - Design and Analysis of Algorithms
- 23DS510 - Numerical Methods
- 23DS520 - Optimization Techniques

**Course Outcomes**

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

CO1	Design randomized algorithms for effectively solving optimization problems.	Apply
CO2	Implement key design paradigms for randomized algorithms.	Apply
CO3	Formulate representative randomized algorithms.	Apply
CO4	Apply randomized and approximation algorithms to combinatorial optimization.	Apply
CO5	Implement LP-based and semi-definite programming methods for optimization.	Apply
CO6	Utilize practical skills through the implementation of randomized algorithms.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	S	L	-	-	-	-	L	L	-	L	S	M
CO2	S	S	S	L	-	-	-	-	-	-	-	-	S	M
CO3	S	S	S	L	-	-	-	-	L	L	-	L	S	M
CO4	S	S	S	L	-	-	-	-	-	-	-	-	S	M
CO5	S	S	S	L	-	-	-	-	-	-	-	-	S	M
CO6	S	S	S	L	-	-	-	-	L	L	-	L	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	5	10	20						25																2	5	15			
CO2	5	10	10						25																2		10			
CO3	5	10	25						50																2	5	15			
CO4											5	10	20								25				2		10			
CO5											5	10	10								25				2		10			
CO6											5	10	25								50				5	15				
Total (in %)	100						100						100						100						100					

## Syllabus

**INTRODUCTION TO RANDOMIZED ALGORITHMS:** Review on Algebra, Number Theory, Combinatorics, and Probability Theory - Randomness as a Source of Efficiency - Designing a Communication Protocol - Models of Randomized Algorithms - Classification: Las Vegas and Monte Carlo (one-sided error, bounded-error, unbounded-error) - Classification of Randomized Algorithms for Optimization Problems.

**DESIGN PARADIGMS:** Foiling the Adversary - Abundance of Witnesses - Fingerprinting - Random Sampling - Amplification - Random Rounding.

**REPRESENTATIVE ALGORITHMS:** Foiling the Adversary - Universal Hashing - Fingerprinting in Communication Protocols - Verification of Matrix Multiplication - Equivalence of Two Polynomials - Success Amplification and Random Sampling - Min-Cut, Satisfiability, and Repeated Random Sampling - Abundance of Witnesses - Optimization & Random Rounding - Primality Testing - Max-SAT Review - Hybrid Sampling & Rounding - Derandomization Techniques.

**INTRODUCTION TO APPROXIMATION ALGORITHMS:** Review on Complexity Theory - Performance Ratios for Approximation Algorithms - Cardinality Vertex-Cover Problem - Well-characterized Problems and Min-Max Relations - Travelling Salesperson Problem (TSP).

**COMBINATORIAL ALGORITHMS:** Set Cover - Steiner Tree - Multi-way Cut and k-Cut - Bin Packing.

**LP-BASED ALGORITHMS:** LP-Duality - Set Cover via Dual Fitting - Set Cover via Primal-Dual Schema - Rounding Applied to Set Cover - Multi-way Cut - Semi-Definite Programming - Randomized Rounding Algorithm - Guarantee Improvement for MAX2SAT.

**PRACTICAL EXERCISES:** Implementation of Randomized Quick Sort - Finding Solutions for S-T Min-Cut Problem - Implementation of Randomized Selection - Implementation of Treap Data Structure - Problems Using Randomized Hash Table - Implementation of Shortest Path and Fast Min-Cut Algorithms - Implementation of Randomized Primality Testing - Implementing the K-Server Online Algorithms.

## Reference Books

1. Rajeev Motwani and Prabhakar Raghavan, "Randomized Algorithms", Cambridge University Press, 1995 (Updated online edition in 2013).
2. Vijay V. Vazirani – "Approximation Algorithms", Springer Science & Business Media, 2013.
3. Juraj Hromkovic– "Design and Analysis of Randomized Algorithms", First edition, Springer, 2005.
4. David P. Williamson and David B. Shmoys, "The Design of Approximation Algorithms", Cambridge University Press, 2011.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
<b>1</b>	<b>INTRODUCTION TO RANDOMIZED ALGORITHMS</b>	
1.1	Review of Algebra, Number Theory, Combinatorics, and Probability Theory	1
1.2	Randomness as a Source of Efficiency	1
1.3	Designing a Communication Protocol and Models of	1

Module No.	Topic	No. of Periods
	Randomized Algorithms	
1.4	Classification: Las Vegas and Monte Carlo (one-sided error, bounded-error, unbounded-error)	1
1.5	Classification of Randomized Algorithms for Optimization Problems	1
<b>2</b>	<b>DESIGN PARADIGMS</b>	
2.1	Foiling the Adversary	1
2.2	Abundance of Witnesses	1
2.3	Fingerprinting	1
2.4	Random Sampling	1
2.5	Amplification	1
2.6	Random Rounding	1
<b>3</b>	<b>REPRESENTATIVE ALGORITHMS</b>	
3.1	Foiling the Adversary, Universal Hashing	1
3.2	Fingerprinting in Communication Protocols, Verification of Matrix Multiplication	1
3.3	Equivalence of Two Polynomials, Success Amplification and Random Sampling	1
3.4	Min-Cut, Satisfiability, and Repeated Random Sampling	1
3.5	Abundance of Witnesses - Optimization & Random Rounding	1
3.6	Primality Testing - Max-SAT Review	1
3.7	Hybrid Sampling & Rounding - Derandomization Techniques.	1
<b>4</b>	<b>INTRODUCTION TO APPROXIMATION ALGORITHMS</b>	
4.1	Review on Complexity Theory, Performance Ratios for Approximation Algorithms	1
4.2	Cardinality Vertex-Cover Problem	1
4.3	Well-characterized Problems and Min-Max Relations	1
4.4	Travelling Salesperson Problem (TSP)	1
<b>5</b>	<b>COMBINATORIAL ALGORITHMS</b>	
5.1	Set Cover	1
5.2	Steiner Tree	1
5.3	Multi-way Cut and k-Cut	1
5.4	Bin Packing	1

Module No.	Topic	No. of Periods
<b>6</b>	<b>LP-BASED ALGORITHMS</b>	
6.1	LP-Duality - Set Cover via Dual Fitting	1
6.2	Set Cover via Primal-Dual Schema - Rounding Applied to Set Cover	1
6.3	Multi-way Cut - Semi-Definite Programming	1
6.4	Randomized Rounding Algorithm - Guarantee Improvement for MAX2SAT.	1
<b>7</b>	<b>PRACTICAL EXERCISES</b>	
7.1	Implementation of Randomized Quick Sort - Finding Solutions for S-T Min-Cut Problem	1
7.2	Implementation of Randomized Selection - Implementation of Treap Data Structure.	1
7.3	Problems Using Randomized Hash Table - Implementation of Shortest Path and Fast Min-Cut Algorithms.	1
7.4	Implementation of Randomized Primality Testing - Implementing the K-Server Online Algorithms	1
	<b>Total</b>	<b>34</b>

### Course Designer

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**23DSPT0 PRINCIPLES OF MANAGEMENT**

Category L T P Credit  
PEC 3 0 0 3

**Preamble**

This course aims to enable students to understand management principles, processes and procedures in consideration of their effort on individual actions.

**Prerequisite**

- Nil

**Course Outcomes**

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

CO1	Explain the fundamental concepts and principles of management.	Understand
CO2	Apply theories to improve the practice of management.	Apply
CO3	Demonstrate the key competencies needed to be an effective manager.	Apply
CO4	Apply managerial ethics and social responsibility in decision-making and organizational practices.	Apply
CO5	Apply an understanding of the complexities in real-life organizations and management situations.	Apply
CO6	Demonstrate critical thinking when presented with managerial issues and problems.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	M	M	M	L	-	-	-	-	-	-	-	-	-
CO2	M	M	M	M	L	-	-	-	-	-	-	-	-	-
CO3	M	M	M	M	M	L	-	M	-	M	S	S	L	S
CO4	M	M	L	M	M	L	L	M	-	M	S	S	L	S
CO5	S	M	S	S	S	L	L	M	S	M	S	S	L	S
CO6	S	S	S	S	S	S	L	S	S	M	S	S	L	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	5	5																							4	6				
CO2	5	15	20																						4	6	10			
CO3	5	15	20																						4	6	10			
CO4			10						50				5	10	15						30				4		10			
CO5									50				5	15	15						40				4	6	10			
CO6													5	15	15						30					6	10			
Total (in %)	100						100						100						100						100					

## Syllabus

**PRINCIPLES OF MANAGEMENT:** Meaning, Definition and Significance of Management, Basic Functions of Management – Planning, Organizing, Staffing, Directing and Controlling. Organizational Environment – Social, Economic, Technological and Political. Corporate Social Responsibility - Case discussion.

**INDUSTRIAL AND BUSINESS ORGANIZATION:** Growth of Industries (Small Scale, Medium Scale and Large Scale Industries). Forms of Business Organizations. Resource Management – Internal and External Sources.

**GROUP BEHAVIOUR:** Group dynamics, Group formation and development, group structure and group cohesiveness. Informal organization – Sociometry – Interaction analysis

**GLOBALISATION:** Issues for global competitiveness, proactive and reactive forces of globalization. Cross cultural management – Management of work force diversity.

**HUMAN RESOURCE MANAGEMENT:** Objectives and Functions, Selection and Placement, Training and Development – Conflict management – Stress management - Human resource management in global environment - Human resource information system(HRIS) - Case discussion.

## Reference Books & web resources

1. Harold Koontz, Heinz Weihrich and Ramachandra Aryasri, "Essentials of Management", Tata McGraw Hill, 11<sup>th</sup> Edition.
2. Ramachandra Aryasri, Harold Koontz and Heinz Weihrich, "Principles of Management" Tata McGraw Hill, 2016, 2<sup>nd</sup> Edition.
3. Mamoria C B and S.V.Gankar, "Personnel Management", 2011.
4. John R Barry, Edward Lee and David G. Messerschmitt, "Digital Communication", 3<sup>rd</sup> Edition. Springer, 2003.

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>PRINCIPLES OF MANAGEMENT</b>	
1.1	Meaning, Definition and Significance of Management,	2
1.2	Basic Functions of Management – Planning, Organizing, Staffing, Directing and Controlling.	2
1.3	Organizational Environment – Social, Economic, Technological and Political.	2
1.4	Corporate Social Responsibility - Case discussion.	2
2	<b>INDUSTRIAL AND BUSINESS ORGANIZATION:</b>	
2.1	Growth of Industries (Small Scale, Medium Scale and Large Scale Industries).	3
2.2	Forms of Business Organizations.	2
2.3	Resource Management – Internal and External Sources.	2
3	<b>GROUP BEHAVIOUR</b>	

3.1	Group dynamics, Group formation and development, group structure and group cohesiveness.	2
3.2	Informal organization – Sociometry – Interaction analysis	3
4	<b>GLOBALISATION</b>	
4.1	Issues for global competitiveness, proactive and reactive forces of globalization.	2
4.2	Cross cultural management – Management of work force diversity.	2
5	<b>HUMAN RESOURCE MANAGEMENT</b>	
5.1	Objectives and Functions, Selection and Placement, Training and Development	2
5.2	Conflict management	2
5.3	Stress management	2
5.4	Human resource management in global environment	2
5.5	Human resource information system(HRIS) - Case discussion.	2
	<b>Total</b>	<b>34</b>

**Course Designer:**

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23DSPU0

**ACCOUNTING AND FINANCIAL  
MANAGEMENT**Category L T P Credit  
PEC 3 0 0 3**Preamble**

This course aims in introducing concepts of accounting and enables an engineer in taking useful financial and costing related decisions by providing scientific tools and techniques.

**Prerequisite**

- NIL

**Course Outcomes**

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

CO1	Explain the basic concepts and process of accounting and finance.	Understand
CO2	Develop trail balance and financial statements like Trading, Profit and Loss accounts, Balance sheet and Cost sheet.	Apply
CO3	Demonstrate the concepts and operations of budgetary control.	Understand
CO4	Apply techniques like break event analysis and budgeting for an organization.	Apply
CO5	Select the right sources of finance and mobilize the right quantum of finance and make use of them in most profitable investment avenues.	Apply
CO6	Apply the concepts if accounting in the real time applications.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	L	L	M	M	L	-	-	-	-	-	-	-	L	M
CO2	L	L	M	M	L	-	-	-	-	-	-	-	L	M
CO3	M	M	M	M	L	-	-	-	-	-	-	-	M	M
CO4	M	M	L	M	L	L	L	L	L	L	L	L	M	S
CO5	S	S	M	M	L	L	L	L	L	M	M	M	S	S
CO6	S	S	S	M	L	L	L	L	L	L	L	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
CO1	5	15																							4	6				
CO2	5	15	20						50																4	6	10			
CO3	5	15																							4	6				
CO4			20						50				5	10											4	6	10			
CO5													10	10	20						50				4	6	10			
CO6													5	20	20						50				4	6	10			
Total (in %)	100						100						100						100						100					

## Syllabus

**Accounting** - Accounting process and principles, financial, cost and management accounting. Elements of book keeping, Journal, cash and handbook, Book reconciliation statement, Ledger, trial balance, profit and loss accounts, final accounts of proprietary and partnership concern and balance sheet.

**Cost accounting** – Objectives, elements of cost, understanding of the different methods of costing.

**Financial Management** – Meaning, scope and role, a brief study of functional areas of financial management. Introduction to various FM tools: Ration Analysis - Meaning - Basis of comparison - Types of ratios

**Working Capital Management** - Theory of Working Capital Management, Introduction, Nature of Working Capital, Concepts and Definitions of Working Capital, Need for Working Capital, Permanent and Temporary Working Capital, Changes in Working Capital, Determinants of Working Capital.

**Budgeting** – budgets, purpose, budgetary control, preparation of budgets, master budget, fixed and flexible budgeting.

## Reference Books

1. M.C.Sukhla, T.S.Grewal, — "Advanced Accounts" Vol III, S.Chand and Publications, New Delhi, 2012
2. Prasanna Chandra, — "Fundamentals of Financial Management", Tata McGraw Hill, 2012.
3. Pandey, —"Financial Management", Vikas Publishing House Pvt. Ltd., 2010
4. Choudhari, Chopde, "Book Keeping and Accountancy"
5. Choudhari, Chopde, "Cost Accounting"
6. M.Y.Khan, P.K. Jain, "Financial Management: Text and Problems"
7. Prasanna Chandra, "Financial Management Theory & Practice", Tata McGraw Hill.
8. Siddiqui S.A. Siddiqui A.S. New Age, "Managerial Economics & Financial Analysis"

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>ACCOUNTING</b>	
1.1	Accounting process and principles, financial, cost and management accounting	2
1.2	Elements of book keeping, Journal, cash and handbook,	2
1.3	Book reconciliation statement, Ledger,	2
1.4	trial balance, profit and loss accounts,	3
1.5	final accounts of proprietary and partnership concern and balance sheet.	3
2	<b>COST ACCOUNTING</b>	
2.1	Objectives, elements of cost, understanding of the different methods of costing.	2
3	<b>FINANCIAL MANAGEMENT</b>	

3.1	Meaning, scope and role, a brief study of functional areas of financial management.	2
4	<b>INTRODUCTION TO VARIOUS FM TOOLS</b>	
4.1	Ration Analysis - Meaning - Basis of comparison - Types of ratios	2
5	<b>WORKING CAPITAL MANAGEMENT</b>	
5.1	Theory of Working Capital Management- Introduction, Nature of Working Capital, Concepts and Definitions of Working Capital	3
5.2	Need for Working Capital, Permanent and Temporary Working Capital,	3
5.3	Changes in Working Capital, Determinants of Working Capital.	3
6	<b>BUDGETING</b>	
6.1	Budgets, purpose, budgetary control	2
6.2	Preparation of budgets, master budget	3
6.3	Fixed and flexible budgeting.	3
	<b>Total</b>	<b>35</b>

**Course Designer(s):**

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**23DSPV0****WIRELESS NETWORK**

Category	L	T	P	Credit
PEC	3	0	0	3

**Preamble**

This course explores the fundamental principles and emerging technologies in wireless communication networks, including protocols, architectures, and standards. It equips students with the skills to analyze, apply, and innovate within the rapidly evolving landscape of wireless networks and mobile communications.

**Prerequisite**

- 23DS550: Computer Networks

**Course Outcomes**

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

CO1	Explain the fundamental concepts of wireless communication, cellular networks, wireless LANs, spectrum allocation, and multiple access methods.	Understand
CO2	Describe the architecture and standards of wireless networks, security protocols, QoS, and related wireless technologies	Understand
CO3	Apply routing protocols in ad hoc and wireless sensor networks, and handle data aggregation, synchronization.	Apply
CO4	Apply mobile network and transport layer protocols to optimize communication in wireless and 2.5/3G networks.	Apply
CO5	Apply physical and MAC layer protocols of Wireless PANs and MANs, to address deployment issues.	Apply
CO6	Apply emerging WLAN technologies, cellular network advancements, and standards to implement solutions for RFID with IoT, M2M communications, and cloud-based services.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	S	L	S	S	M	-	-	-	M	-	-	L	M
CO2	M	S	M	S	M	-	-	-	-	M	-	-	L	M
CO3	L	S	L	S	S	-	-	-	-	-	-	-	M	S
CO4	M	S	L	S	S	-	-	-	-	-	-	-	L	M
CO5	L	L	S	S	S	-	-	-	-	-	-	-	L	M
CO6	M	S	L	S	S	M	-	-	-	-	-	-	M	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's Level																														
CO1	10	10					25																		3	10				
CO2	10	15					25																		3	10				
CO3	10	15	30					50																	3		15			
CO4													5	10	10										3		15			
CO5													5	10	20										3		15			
CO6													10	10	20											5	15			
Total (in %)	100						100						100						100											

## Syllabus

**WIRELESS FUNDAMENTALS:** Introduction to cellular networks,-wireless local area networks- Spectrum allocations – Radio propagation models-Narrowband digital modulation and wireless fading environments. – Modern Communications Systems – MAC:SDMA – TDMA – FDMA - CDMA

**WLAN TECHNOLOGIES:** wireless network architectures – 802.11 Physical (PHY) layers – 802.11 MAC – Security Protocols: WPA WPA2, WPA3, IEEE 802.11i, Quality of Service (QoS): MAC enhancements: IEEE 802.11e– Related Wireless Standards (Hyperlan, HomeRF, Bluetooth, Zigbee, Wireless USB)- WiFi (IEEE 802.11x) and WiMAX (IEEE 802.16x) Standards

**AD HOC AND SENSOR NETWORKS:** Ad hoc Network- Characteristics- Table-driven (Proactive) and Source-initiated On Demand routing protocols (Reactive), Hybrid protocols - Routing in intermittently connected mobile networks. **Wireless Sensor networks-** Classification, MAC and Routing Protocols, Data Aggregation and Synchronization.

**MOBILE NETWORK AND TRANSPORT LAYERS:** Mobile IP – Dynamic Host Configuration Protocol-Mobile Ad Hoc Routing Protocols (AODV,DSR, OLSR)–Multicast routing protocols in wireless networks -TCP over Wireless Networks – Indirect TCP – Snooping TCP – MobileTCP – Transaction Oriented TCP- TCP over 2.5 / 3G wireless Networks.

**WIRELESS PANs, MANs** – Physical and MAC layer details, Wireless PANs – Architecture of Bluetooth Systems, Physical and MAC layer details, Standards- WLAN deployment issues- Interference – Resource Allocation, Wireless MANs (WiMAX, IEEE 802.16) architecture and standards

**FUTURE TRENDS:** Emerging WLAN Related Technologies – (IEEE 802.11ac, 802.11ax) Trends – Cellular network evolution– IEEE 802.20 (Mobile Broadband Wireless Access), IEEE 802.22 (Wireless Regional Area Networks), UWB, Cognitive Radio Networks, RFID – Convergence of 4G/5G with IoT, M2M communications, and cloud-based services

## Reference Books & web resource

1. Gary. S. Rogers and John Edwards, “An Introduction to Wireless Technology”, Pearson Education, 2012.
2. SivaRam Murthy C and B.S Manoj, “Ad hoc Wireless Networks Architecture and Protocols”, Pearson Education, 2005.
3. KavehPahlavan, Prashant K. Krishnamurthy, “Principles of wireless networks : A unified approach”, John Wiley & Sons, 2011.
4. William Stallings, “Wireless Communication and Networks”, Pearson Education, 2009.
5. Dharma Prakash Agrawal and Qing-An Zeng, “Introduction to Wireless and Mobile Systems”, Thomson Press, 2007.
6. “Ad Hoc Networks: Fundamental Properties and Network Topologies" by Junshan Zhang (2021)
7. "Wireless Communications" by Andreas F. Molisch (3rd Edition, 2022)
8. **NPTEL Course:** Introduction to Wireless and Cellular Communications  
<https://archive.nptel.ac.in/noc/courses/noc17/SEM2/noc17-cs37/>

## Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	WIRELESS FUNDAMENTALS	

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
1.1	Introduction to cellular networks	1
1.2	wireless local area networks- Spectrum allocations	1
1.3	Radio propagation models	1
1.4	Narrowband digital modulation and wireless fading environments	1
1.5	Modern Communications Systems – Introduction to MAC	1
1.6	SDMA – TDMA -FDMA - CDMA	2
1.7	Cellular and Ad-hoc-Concepts.	1
<b>2</b>	<b>WLAN TECHNOLOGIES</b>	
2.1	wireless network architectures	1
2.2	802.11 Physical (PHY) layers – 802.11 MAC	1
2.3	Security Protocols: WPA WPA2, WPA3, IEEE 802.11i,	1
2.4	Quality of Service (QoS): MAC enhancements: IEEE 802.11e– Related Wireless Standards	1
2.5	WiFi (IEEE 802.11x) and WiMAX (IEEE 802.16x) Standards	1
<b>3</b>	<b>AD HOC AND SENSOR NETWORKS</b>	
3.1	Ad hoc Network- Characteristics- Table-driven (Proactive, Reactive routing protocols)	1
3.2	Hybrid protocols - Routing in intermittently connected mobile networks.	1
3.3	Wireless Sensor networks- Classification, MAC and Routing Protocols,	1
3.4	Data Aggregation and Synchronization.	1
<b>4</b>	<b>MOBILE NETWORK AND TRANSPORT LAYERS:</b>	
4.1	Mobile IP – Dynamic Host Configuration Protocol-	1
4.2	Mobile Ad Hoc Routing Protocols (AODV,DSR, OLSR)	1
4.3	TCP over Wireless Networks	1
4.4	Indirect TCP – Snooping TCP – MobileTCP	1
4.5	Fast Retransmit / Fast Recovery	1
4.6	Transmission/Timeout Freezing	1
4.7	Selective Retransmission	1
4.8	TCP over 2.5 / 3G wireless Networks.	1
<b>5</b>	<b>WIRELESS PANs MANs</b>	
5.1	Physical and MAC layer details	1

Module No.	Topic	No. of Periods
5.2	Wireless PANs	1
5.3	Architecture of Bluetooth Systems	1
5.4	Physical and MAC layer details, Standards	1
5.5	WLAN deployment issues	1
5.6	Interference – Resource Allocation,	1
5.7	Wireless MANs (WiMAX, IEEE 802.16) architecture and standards	1
<b>6</b>	<b>FUTURE TRENDS</b>	
6.1	Emerging WLAN Related Technologies – (IEEE 802.11ac, 802.11ax) Trends	1
6.2	Cellular network evolution (LTE, LTE-Advanced, 5G, 6G)	1
6.3	IEEE 802.20 (Mobile Broadband Wireless Access), IEEE 802.22 (Wireless Regional Area Networks),	1
6.4	UWB, Cognitive Radio Networks, RFID, Convergence of 4G/5G with IoT	1
	<b>Total</b>	<b>36</b>

**Course Designer:**

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**23DSPW0****NETWORKSCIENCE**

Category L T P Credit

PEC 3 0 0 3

**Preamble**

This course explores the identification, construction, visualization, and analysis of networks using mathematical and algorithmic approaches. Students will gain the ability to explain key concepts and results in network science both mathematically and conceptually.

**Prerequisite**

- 23DS220: Applied Statistics
- 23DS230 Graph Theory
- 23DS550: Computer Networks

**Course Outcomes**

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

CO1	Explain the fundamental concepts of graph theory, including networks, random networks, degree distributions, and clustering coefficients.	Understand
CO2	Describe the concepts of random and scale-free networks, including degree distribution, network evolution, and the role of hubs	Understand
CO3	Apply the Barabási-Albert model to simulate network growth through preferential attachment,	Apply
CO4	Examine the usage of Bianconi-Barabási model and bose-einstein condensation in network working	Apply
CO5	Apply methods for measuring degree correlations, such as assortativity and disassortativity, to identify patterns in real-world networks.	Apply
CO6	Analyze the robustness and vulnerability of networks using percolation theory, identifying critical nodes and links in a network.	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	S	S	M	L	-	L	-	-	-	-	L	S	M
CO2	S	S	S	M	L	-	L	-	-	-	-	L	S	M
CO3	S	S	S	S	L	-	M	-	-	-	-	L	S	S
CO4	S	S	S	S	L	-	M	-	-	-	-	L	S	S
CO5	S	S	S	S	L	-	M	-	-	-	-	L	S	S
CO6	S	S	S	M	L	-	M	-	-	-	-	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						Assignment 1						CAT2						Assignment2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's Level																														
CO1	10	15					25																		3	10				
CO2	10	15					25																		3	10				
CO3	10	15	25					50																	3		15			
CO4													5	10	10										3		15			
CO5													5	10	10										3		15			
CO6														10	15	25											40			
Total (in %)	100						100						100						100											

**Syllabus**

**GRAPH THEORY**-Networks and graphs, Degree, Average degree and Degree distribution, Real networks are sparse, Adjacency Matrix, Weighted and unweighted networks, Bipartite networks, Paths and distances in networks, Connectedness and Components, Clustering coefficient.

**RANDOM NETWORKS**- The random network model, number of links variable, Degree distribution, Evolution of Random network, Six degrees of separation.

**Scale free Property**- Power laws and scale-free networks, Hubs, Universality, Ultra-small property, Role of the degree exponent, Generating networks with a pre-defined degree distribution

**BARABÁSI-ALBERT MODEL**- Growth and preferential attachment, Barabási-Albert Model, Degree dynamics, Degree distribution, The absence of growth or preferential attachment, Non-linear preferential attachment.

**EVOLVING NETWORKS**- The Bianconi-Barabási model, Measuring fitness, Bose-Einstein condensation, Evolving Networks

**DEGREE CORRELATIONS**- Assortativity and disassortativity, Measuring degree correlations, Structural cutoffs, Degree correlations in real networks, Generating correlated networks

**NETWORK ROBUSTNESS**: Percolation theory, robustness of scale-free networks, attack tolerance, cascading failures, modelling cascading failures, building robustness.

**Reference Books & web resources**

1. Albert-Laszlo Barabasi, "Network Science", 1st Edition, Cambridge Univ. Press, 2016.
2. Ted G. Lewis, "Network Science: Theory and Practice", Wiley, 2013
3. Estrada, E., Fox, M., Higham, D.J. and Oppo, G.L.,- "Network Science - Complexity in Nature and Technology", Springer,2010.

<https://eravilaipnada.wordpress.com/wp-content/uploads/2016/01/barabasi-2012.pdf>

**Course Contents and Lecture Schedule**

Module No.	Topic	No. of Periods
<b>1</b>	<b>GRAPH THEORY</b>	
1.1	Networks and graphs - introduction	1
1.2	Degree, Average degree and Degree distribution,	1
1.3	Real networks are sparse, Adjacency Matrix	1
1.4	Weighted and unweighted networks, Bipartite networks,	2
1.5	Paths and distances in networks, Connectedness and Components, Clustering coefficient.	2
<b>2</b>	<b>RANDOM NETWORKS</b>	
2.1	The random network model, The number of links is variable,	1
2.2	Degree distribution	1

Module No.	Topic	No. of Periods
2.3	Evolution of Random network, Six degrees of separation.	1
<b>3</b>	<b>SCALE FREE PROPERTY</b>	
3.1	Power laws and scale-free networks	1
3.2	Hubs, Universality, Ultra-small property Role of the degree exponent	2
3.3	Generating networks with a pre-defined degree distribution	2
<b>4</b>	<b>BARABÁSI-ALBERT MODEL</b>	
4.1	Growth and preferential attachment	1
4.2	Barabási-Albert Model	2
4.3	Degree dynamics, Degree distribution	2
4.4	The absence of growth or preferential attachment,	1
4.5	Non-linear preferential attachment	1
<b>5</b>	<b>EVOLVING NETWORKS</b>	
5.1	The Bianconi-Barabási model	2
5.2	Measuring fitness, Bose-Einstein condensation	2
5.3	Evolving Networks	1
<b>6</b>	<b>DEGREE CORRELATIONS</b>	
6.1	Assortativity and disassortativity	1
6.2	Measuring degree correlations, Structural cutoffs,	1
6.3	Degree correlations in real networks	1
6.4	Generating correlated networks	1
<b>7</b>	<b>NETWORK ROBUSTNESS</b>	
7.1	Percolation theory, robustness of scale-free networks	1
7.2	Attack tolerance, cascading failures	1
7.3	Modelling cascading failures, building robustness	1
	<b>Total</b>	<b>34</b>

**Course Designer(s):**

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**23DSPX0 INFORMATION RETRIEVAL**

Category L T P Credit  
PEC 3 0 0 3

**Preamble**

The course explores foundational and advanced concepts in Information Retrieval, including search models, indexing, and evaluation techniques. Students will implement real-world IR systems, study modern techniques like neural models and knowledge graphs, and address ethical concerns such as bias and privacy in search systems.

**Prerequisite**

- 23DS330 – Data Structures and Algorithms
- 23DS210 23DS220, – Probability and Statistics
- 23DS540 - Machine Learning

**Course Outcomes**

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

CO1	Understand the fundamental concepts of IR, including document representation, retrieval models, and evaluation metrics	Understand
CO2	Apply indexing, vector-space models, and probabilistic retrieval methods to build and test IR systems.	Apply
CO3	Analyze the efficiency and effectiveness of IR models using real-world data.	Analyze
CO4	Apply algorithms for modern web search engines, including PageRank and machine learning-based ranking	Apply
CO5	Apply the ethical and societal impacts of search algorithms, particularly issues like privacy, bias, and fake news	Apply
CO6	Analyze and implement a basic information retrieval system that can handle large-scale document collections	Analyze

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	L	M	L	L	L	-	-	L	-	-	-	M	M	S
CO2	L	M	S	M	L	-	-	L	-	-	-	M	M	M
CO3	L	S	L	S	S	-	-	L	-	-	-	M	S	S
CO4	L	S	S	L	S	-	-	M	-	-	-	S	S	S
CO5	L	M	M	L	L	-	-	S	-	-	-	S	S	M
CO6	L	S	S	S	S	-	-	L	-	-	-	M	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT 1						Assignment 1						CAT 2						Assignment 2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																														
CO1	5	20																							3	10				
CO2	5	10	20						50																3		10			
CO3		10	10	20						50															3	5		15		
CO4													5	10	20									3	5	10				
CO5													5	10	20					50				3		10				
CO6															10	20						50			5		15			
Total (in %)	100						100						100						100											

## Syllabus

**INTRODUCTION TO INFORMATION RETRIEVAL** - History and Components of Information Retrieval: Overview of the history of IR systems, Components of IR systems (document representation, indexing, and retrieval) - Open-source IR Frameworks: Popular open-source IR systems (Lucene), Impact of the web on IR - Boolean Retrieval Model: Basics of Boolean retrieval, Boolean operators and query formulation, Advantages and limitations - Vector Space Model: Vector space representation of documents, Term weighting and TF-IDF, Cosine similarity for ranking documents - Evaluation of IR Systems: Evaluation metrics: Precision, Recall, F-measure, Normalized Discounted Cumulative Gain (nDCG), Relevance feedback and query expansion.

**INDEXING AND RETRIEVAL METHODS** - Inverted Indices: Construction of an inverted index, Efficient processing of sparse vectors, Compressed indices – Pre - processing Techniques: Tokenization, stopword removal, stemming, and lemmatization - Language Models for IR: Basics of Language Model-based IR, Probabilistic IR and the binary independence model, Language models vs other approaches to IR - Relevance Feedback: Concept of relevance feedback, Techniques for relevance feedback and query reformulation - Web Search Basics: Introduction to web search, Crawling and indexing web pages, Index compression and evaluation.

**WEB SEARCH AND MODERN RETRIEVAL:** PageRank and Link Analysis: Hyperlink-induced topic search (HITS) algorithm, PageRank algorithm, Applications of link analysis in web search - Search Engine Optimization (SEO) and Web Structure: Introduction to SEO and its importance, Measuring web size and structure - Focused Crawling and Meta-Crawlers: Focused crawling techniques - Relevance Scoring and Ranking in Web Search: Ranking methods for web search, Scoring documents based on term importance - Personalized Search and Collaborative Filtering: Techniques for personalized search, Recommender systems: collaborative filtering and content-based filtering - Handling the “Invisible Web: Challenges in crawling the deep web, Techniques to retrieve information from the deep/invisible web - Hadoop and MapReduce for Large-Scale Retrieval: Introduction to Hadoop, Using MapReduce for processing large datasets.

**ADVANCED RETRIEVAL MODELS** - Latent Semantic Indexing (LSI): Singular Value Decomposition (SVD), Use of LSI in IR systems, Comparison with vector-space models - Neural Ranking Models: Introduction to neural information retrieval, Comparison of dense vs sparse retrieval models - Knowledge Graphs and Semantic Search: Role of knowledge graphs in modern IR systems, Semantic search using knowledge graphs, Applications in question answering systems.

**RESPONSIBLE IR** - Bias, Fairness, and Ethics in IR Systems: Understanding bias in search algorithms, Fairness and responsible use of IR technology, Ethical considerations in search - Fake News and Privacy in Search Engines: Identifying and handling fake news, Ensuring user privacy in IR systems

## Reference Books & web resources

1. Christopher D. Manning, Prabhakar Raghavan, Hinrich Schütze, “Introduction to Information Retrieval”, Cambridge University Press, 2009.

2. Ricardo Baeza-Yates, Berthier Ribeiro-Neto, "Modern Information Retrieval: The Concepts and Technology behind Search", 2nd Edition, ACM Press Books, 2011.
3. Stefan Büttcher, Charles L. A. Clarke, Gordon V. Cormack, "Information Retrieval: Implementing and Evaluating Search Engines", MIT Press, 2010.
4. Mark Levene, "An Introduction to Search Engines and Web Navigation", 2nd Edition, Wiley, 2010.

### Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>INTRODUCTION TO INFORMATION RETRIEVAL</b>	
1.1	History and Components of Information Retrieval: Overview of the history of IR systems, Components of IR systems (document representation, indexing, and retrieval)	1
1.2	Open-source IR Frameworks: Popular open-source IR systems (Lucene), Impact of the web on IR	1
1.3	Boolean Retrieval Model: Basics of Boolean retrieval, Boolean operators and query formulation, Advantages and limitations	2
1.4	Vector Space Model: Vector space representation of documents, Term weighting and TF-IDF, Cosine similarity for ranking documents	2
1.5	Evaluation of IR Systems: Evaluation metrics: Precision, Recall, F-measure, Normalized Discounted Cumulative Gain (nDCG), Relevance feedback and query expansion	3
2	<b>INDEXING AND RETRIEVAL METHODS</b>	
2.1	Inverted Indices: Construction of an inverted index, Efficient processing of sparse vectors, Compressed indices	2
2.2	Preprocessing Techniques: Tokenization, stopword removal, stemming, and lemmatization	1
2.3	Language Models for IR: Basics of Language Model-based IR, Probabilistic IR and the binary independence model, Language models vs other approaches to IR	2
2.4	Relevance Feedback: Concept of relevance feedback, Techniques for relevance feedback and query reformulation	1
2.5	Web Search Basics: Introduction to web search, Crawling and indexing web pages, Index compression and evaluation	2
3	<b>WEB SEARCH AND MODERN RETRIEVAL</b>	
3.1	PageRank and Link Analysis: Hyperlink-induced topic search (HITS) algorithm, PageRank algorithm, Applications of link analysis in web search	1
3.2	Search Engine Optimization (SEO) and Web Structure: Introduction to SEO and its importance, Measuring web size and structure	1
3.3	Focused Crawling and Meta-Crawlers: Focused crawling techniques	1
3.4	Relevance Scoring and Ranking in Web Search: Ranking methods for web search, Scoring documents based on term importance	2
3.5	Personalized Search and Collaborative Filtering: Techniques for	2

	personalized search, Recommender systems: collaborative filtering and content-based filtering	
3.6	Handling the “Invisible Web: Challenges in crawling the deep web, Techniques to retrieve information from the deep/invisible web	1
3.7	Hadoop and MapReduce for Large-Scale Retrieval: Introduction to Hadoop, Using MapReduce for processing large datasets	1
4	<b>ADVANCED RETRIEVAL MODELS</b>	
4.1	Latent Semantic Indexing (LSI): Singular Value Decomposition (SVD), Use of LSI in IR systems, Comparison with vector-space models	2
4.2	Neural Ranking Models: Introduction to neural information retrieval, Comparison of dense vs sparse retrieval models	2
4.3	Knowledge Graphs and Semantic Search: Role of knowledge graphs in modern IR systems, Semantic search using knowledge graphs, Applications in question answering systems	2
5	<b>RESPONSIBLE IR</b>	
5.1	Bias, Fairness, and Ethics in IR Systems: Understanding bias in search algorithms, Fairness and responsible use of IR technology, Ethical considerations in search	2
5.2	Fake News and Privacy in Search Engines: Identifying and handling fake news, Ensuring user privacy in IR systems	2
	<b>Total</b>	<b>36</b>

**Course Designer(s):**

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**23DS1A0****DATA REPORTING**

Category	L	T	P	Credit
EEC	1	0	0	1

**Preamble**

This course aims at providing introduction to data reporting tools and hands on training to report data in appropriate formats.

**Prerequisite**

- Nil

**Course Outcomes**

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

**CO1:** Explain the basics of data reporting and data integration Understand

**CO2:** Use data reporting tools to generate reports with available data Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	P10	P11	P12	PSO1	PSO2
CO1.	S	M	S	M	M	L	L	L	M	L	L	M	M	S
CO2.	M	S	M	S	L	L	M	M	L	L	M	S	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	Internal						Terminal						
	Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6
CO1		10	10	20				10	10	20			
CO2		10	20	30				10	20	30			
Total (in %)		100						100					

**Syllabus**

Basics of data reporting - Data integration from different sources - Generating reports/ Dashboard - Measuring domain specific KPIs - Sample data and project work  
Tools: Excel, Tableau

**Reference Books**

1. Winston, W., "Microsoft Excel 2013 Data Analysis and Business Modeling: Data Analysis and Business Modeling", Pearson Education, 2014
2. Sleeper, R.,. Practical tableau: 100 tips, tutorials, and strategies from a Tableau zen master. " O'Reilly Media, Inc.", 2018.

**Course Designers:**

1. <Ms. S. Suthalakshmi, Data Scientist, Tiger Analytics> <sutha.lakkshmi@gmail.com>
2. <Dr. D.Anitha> <anithad@tce.edu>

**23DS1B0      TABLEAU IN DATA VISUALIZATION**

Category	L	T	P	Credit
EEC	1	0	0	1

**Preamble**

This course aims at providing introduction to applications of tableau in data visualisation and hands on training to visualise data in diverse formats using tableau.

**Prerequisite**

- Nil

**Course Outcomes**

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

**CO1:** Explain the basics visualization tools of Tableau Understand

**CO2:** Use tableau to generate visualisation reports with available data Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	P10	P11	P12	PSO1	PSO2
CO1	S	M	S	M	M	L	L	L	M	L	L	M	M	S
CO2	M	S	M	S	L	L	M	M	L	L	M	S	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	Internal						Terminal						
	Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6
CO1		10	10	20				10	10	20			
CO2		10	20	30				10	20	30			
Total (in %)		100						100					

**Syllabus**

Basics of Business Intelligence-Introduction to Tableau - Connecting to Data Sources- Querying data from Data Sources- Data wrangling- Understanding the Workspace- Features(Attributes, Dimensions and Calculate fields)- Chart Features-Filters and Marks- Tableau Chart Demonstration Examples-Sample data and project work.

**Reference Books**

1. Marleen Meier, David Baldwin,"Mastering Tableau 2021: Implement advanced business intelligence techniques and analytics with Tableau",Packt Publishing,2021.
2. Knaflic, Cole Nussbaumer,"Storytelling with data a data visualization guide for business professionals",Wiley Publisher,2022.
3. Lindy Ryan,"Visual Data Storytelling with Tableau",Addison-Wesley Professional,2018.

**Course Designers:**

1. <Ms.Christy Britto, Morgan Stanley> <jbchristy@gmail.com>
2. <Ms. R. Saraswathy Meena> <rsmca@tce.edu>

**23DS1C0****NODE JS**

Category	L	T	P	Credit
EEC	1	0	1	1

**Preamble**

This course provides an introduction to Node.js, a powerful JavaScript runtime built on Chrome's V8 engine. Students will learn the core concepts of Node.js, such as its event-driven architecture, non-blocking I/O, and the Node.js execution model. The course also covers the Express.js framework for building web applications, connecting to databases like MongoDB or MySQL, and error handling techniques. Students will engage in practical labs, creating APIs, working with files, and managing processes using PM2.

**Prerequisite**

- 23DS570 - Web Technology Lab

**Course Outcomes**

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

CO1	Apply the fundamental concepts of Node.js, including its event-driven architecture, core modules, and the Node Package Manager	Apply
CO2	Develop advanced Node.js applications using asynchronous programming, Express.js, and database integration	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	P10	P11	P12	PSO1	PSO2
CO1	S	M	S	M	M	L	L	L	M	L	L	M	M	S
CO2	M	S	M	S	L	L	M	M	L	L	M	S	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	Internal						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6
<b>Bloom's level</b>	1	2	3	4	5	6	1	2	3	4	5	6
<b>CO1</b>	10	10	20				10	10	20			
<b>CO2</b>	10	20	30				10	20	30			
<b>Total (in %)</b>	100						100					

**Syllabus****Introduction and Fundamentals**

**Introduction to Node.js:** Basics, advantages, and event-driven architecture.

**Core Concepts of Node.js:** Execution model, asynchronous programming, and non-blocking I/O.

**Modules in Node.js:** Overview of built-in modules like fs, path, and http.

**Introduction to NPM:** Package management, installation, and significance of package.json and package-lock.json.

**Advanced Concepts and Practical Applications**

**Asynchronous Programming:** Callbacks, Promises, and async/await syntax.

**Express.js Framework:** Building web applications, routing, and middleware.

**Working with Databases:** Connecting to MongoDB and MySQL, CRUD operations.

**Error Handling and Debugging:** Error handling strategies in Node.js applications.

**PM2 Process Manager:** Managing Node.js processes and monitoring application performance.

**Lab (Hands-on - Basic)**

**Setting up the Environment:** Installing Node.js and npm; running the first application.

**Modules in Node.js:** Creating and using custom modules; practice with built-in modules.

**Working with the File System:** Reading/writing files using the fs module.

**Working with HTTP:** Creating a basic HTTP server and serving static files.

**Lab (Hands-on - Advanced)**

**Asynchronous Programming:** Building an application using callbacks, Promises, and async/await.

**Express.js Framework:** Setting up an Express app and creating routes.

**Working with Databases:** Performing CRUD operations and managing data.

**Error Handling and Debugging:** Implementing error handling and debugging techniques.

**PM2 Process Manager:** Installing, configuring, and monitoring Node.js apps with PM2.

**Building a Simple API:** Creating RESTful routes, handling JSON data, and using Postman for testing.

**Reference Books**

1. Andrew Mead *"Learning Node.js Development"*, Packt Publishing, 2018.
2. Mario Casciaro, *"Luciano Mammino Node.js Design Patterns"*, Second Edition, Packt Publishing, 2016.
3. Jim R. Wilson, *"Node.js 8 the Right Way: Practical, Server-Side JavaScript That Scales"*, Pragmatic Bookshelf, 2018.
4. Mike Cantelon, Marc Harter, T.J. Holowaychuk, Nathan Rajlich, *"Node.js in Action"*, Second Edition, Manning Publications, 2017.

**Course Contents and Lecture Schedule**

Module No.	Topic	No. of Lectures
1.	<b>Introduction and Fundamentals</b>	
1.1	Introduction to Node.js	1
1.2	Core Concepts of Node.js	1
1.3	Modules in Node.js and Introduction to NPM	1
2.	<b>Advanced Concepts and Practical Applications</b>	
2.1	Asynchronous Programming	1
2.2	Express.js Framework	1
2.3	Error Handling and Debugging	1
2.4	Working with Databases and PM2 Process Manager	1
3.	<b>Lab (Hands-on - Basic)</b>	
3.1	Setting up the Environment and Modules in Node.js	1
3.3	Working with the File System and Working with HTTP	1
4.	<b>Lab (Hands-on - Advanced)</b>	
4.1	Asynchronous Programming	1
4.2	Express.js Framework	1
4.3	Working with Databases	1
4.4	Error Handling and Debugging	1
4.5	PM2 Process Manager and Building a Simple API	1
<b>Total</b>		<b>14</b>

**Course Designers:**

1. **Faculty** : Dr.C.Mahadevi.,Assistant Professor, Applied Mathematics and Computational Science(AMCS) in Data Science, [cmids@tce.edu](mailto:cmids@tce.edu)
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**23DS1D0****PROMPT ENGINEERING**

Category	L	T	P	Credit
EEC	1	0	0	1

**Preamble**

This course provides an introduction to prompt engineering techniques, focusing on crafting effective prompts for AI models. It includes hands-on training for optimizing prompts and evaluating their performance across various applications.

**Prerequisite**

- Nil

**Course Outcomes**

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

**CO1:** Develop and implement effective prompting techniques to optimize AI model responses across various applications Apply

**CO2:** Assess and evaluate the security, alignment, and performance of prompting strategies through practical case studies and benchmarking methodologies. Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1.	S	M	S	M	M	L	L	L	M	L	L	M	M	S
CO2.	M	S	M	S	L	L	M	M	L	L	M	S	S	M

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	Internal						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level												
CO1	10	10	20				10	10	20			
CO2	10	20	30				10	20	30			
Total (in %)	100						100					

**Syllabus**

**Introduction to Prompts:** Key Terminology and Components, Historical Context of Prompting

**Text-Based Techniques:** In-Context Learning, Zero-Shot, Thought Generation

**Advanced Techniques:** Decomposition, Ensembling, Self-Criticism.

**Answer Engineering:** Shape, Space, and Extraction.

**Multilingual Prompting and Techniques Extensions:** Tool Use Agents, Code-Generation Agents.

**Retrieval Augmented Generation Prompting Issues:** Security Risks and Alignment Challenges Benchmarking Techniques - Case Study

**Reference Books**

1. El Amri, A. (2024). LLM Prompt Engineering for Developers. Packt Publishing.
2. Berryman, J., & Ziegler, A. (2024). Prompt Engineering for LLMs. O'Reilly Media.

**Course Contents and Lecture Schedule**

Module No.	Topic	No. of Lectures
<b>1.</b>	<b>Introduction to Prompts</b>	
1.1	Key Terminology and Components	1
1.2	Historical Context of Prompting	1
<b>2.</b>	<b>Text-Based Techniques</b>	
2.1	In-Context Learning, Zero-Shot	1
2.2	Thought Generation	1
<b>3.</b>	<b>Advanced Techniques</b>	
3.1	Decomposition, Ensembling, Self-Criticism	2
<b>4.</b>	<b>Answer Engineering</b>	
4.1	Shape, Space, and Extraction.	2
<b>5.</b>	<b>Multilingual Prompting and Techniques Extensions</b>	
5.1	Tool Use Agents	1
5.2	Code-Generation Agents.	1
<b>6.</b>	<b>Retrieval Augmented Generation Prompting Issues</b>	
6.1	Security Risks and Alignment Challenges Benchmarking Techniques	1
6.2	Case Study	1
<b>Total</b>		<b>12</b>

**Course Designers:**

1. <Mr. James Sabarimuthu> <[james.sabarimuthu@gmail.com](mailto:james.sabarimuthu@gmail.com)>  
Conversight.AI
2. <Dr. S. T. Padmapriya> <[stpca@tce.edu](mailto:stpca@tce.edu)>

23DS1E0

**SMART DATA  
INFRASTRUCTURE**

Category	L	T	P	Credits
EEC	1	0	0	1

**Preamble**

This course focuses on developing smart data infrastructures that seamlessly connect automation, monitoring, and visualization layers into a unified operational ecosystem. It emphasizes the implementation of real-time data collection and monitoring solutions using modern open-source platforms, integrating metrics, logs, and analytics dashboards to enhance operational intelligence and decision-making. Through this approach, learners gain practical experience in building data-driven systems that support efficient analysis, transparency, and continuous performance optimization.

**Prerequisite**

- Nil

**Course Outcomes**

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

CO1	Design and implement smart data pipelines that integrate real-time monitoring, automation, and visualization components using open-source platforms.	Apply
CO2	Construct and utilize metrics, logs, and analytics dashboards to support operational intelligence and informed decision-making.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	M	M	S	S	S	L	L	L	M	M	L	S	S	M
CO2	M	S	M	S	S	L	L	M	M	S	M	S	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	Internal Assessment						Terminal Examination					
	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level												
CO1	10	10	20				10	10	20			
CO2	10	20	30				10	20	30			
Total (in %)	100						100					

**Syllabus**

**Smart Data Infrastructure: Concepts and Foundations** - Overview of Smart Data Systems and Infrastructure Layers - Components of Modern Data Pipelines: Collection, Processing, and Visualization - Overview to Workflow Orchestration and Observability - Architecture of Monitoring Systems: Metrics, Logs, and Alerts.

**Workflow Integration and Real-Time Visualization:** Building Connected Workflows for Data Collection and Updates - Configuring Metric Collection using **Prometheus** - Real-Time Visualization Dashboards using **Grafana** - Integrating Automated Alerts and Threshold Events.

**Text Book**

1. Kleppmann, Martin. *Designing Data-Intensive Applications*. O'Reilly Media, 2017.
2. Villamizar, S. *Data Observability and Monitoring for Modern Data Systems*. Apress, 2023.

### Reference Books & web resources

1. Grafana Labs Documentation – <https://grafana.com/docs/>
2. Prometheus Official Docs – <https://prometheus.io/docs/>

### Course Contents and Lecture Schedule

Module No.	Topic	No. of Periods
1	<b>Smart Data Infrastructure: Concepts and Foundations</b>	
	Overview of Smart Data Systems and Infrastructure Layers	1
	Components of Modern Data Pipelines: Collection, Processing, and Visualization	2
	Introduction to Workflow Orchestration and Observability	2
	Architecture of Monitoring Systems: Metrics, Logs, and Alerts	2
2	<b>Workflow Integration and Real-Time Visualization</b>	
	Building Connected Workflows for Data Collection and Updates	2
	Configuring Metric Collection using <b>Prometheus</b>	1
	Real-Time Visualization Dashboards using <b>Grafana</b>	2
	Integrating Automated Alerts and Threshold Events	2
	Total	14

### Course Designer(s):

1. Mohamed Arsath S, Director of Operations, Tarcin [director@tarcin.in](mailto:director@tarcin.in)
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<b>23DS2F0</b>	<b>AI-POWERED ENTERPRISE DATA MANAGEMENT</b>	Category	L	T	P	Credits
		EEC	2	0	0	2

**Preamble**

Modern enterprises generate large volumes of data through ERP systems such as SAP, Oracle, Salesforce, and Microsoft Dynamics, making effective data management essential for analytics, digital transformation, and AI-driven decision-making. This short-duration course introduces the fundamentals of enterprise data management using AI-assisted tools. It provides a step-by-step understanding of the enterprise data lifecycle, including data assessment, extraction, transformation, loading, and governance, and highlights how AI technologies can enhance and automate enterprise data management processes.

**Prerequisite**

- Database concepts and SQL
- Data processing fundamentals
- Programming basics (Python preferred)
- Basic understanding of data science or analytics

**Course Outcomes**

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

CO1	Explain the enterprise data ecosystem and ERP data structures.	Apply
CO2	Perform basic data assessment and extraction using AI-assisted techniques.	Apply
CO3	Apply data transformation, cleansing, and validation methods to improve data quality.	Apply
CO4	Apply enterprise data loading, reconciliation, and governance practices to ensure data quality.	Apply

**Mapping with Programme Outcomes**

COs	PO1	PO 2	PO 3	PO 4	PO 5	PO 6	PO 7	PO 8	PO 9	PO1 0	PO1 1	PO1 2	PSO 1	PSO 2
CO1	M	M	-	-	S	-	-	-	-	M	-	-	-	M
CO2	M	M	-	S	S	-	-	-	-	-	-	-	M	S
CO3	M	M	M	S	S	-	-	-	-	-	-	-	S	S
CO4	M	M	M	S	S	-	-	M	-	-	-	-	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	Internal						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level	1	2	3	4	5	6	1	2	3	4	5	6
CO1	4	10	10				4	10	10			
CO2	4	10	10				4	10	10			
CO3	4	10	12				4	10	12			
CO4	4	10	12				4	10	12			
<b>Total (in %)</b>	<b>100</b>						<b>100</b>					

## Syllabus

**Enterprise Data Ecosystem & ERP Fundamentals:** Introduction to enterprise information systems. Overview of ERP platforms such as SAP, Oracle, Salesforce, and MS Dynamics. Understanding enterprise data categories including master data, transactional data, and metadata. Structure of enterprise databases and data relationships. Introduction to enterprise data migration lifecycle and the role of AI in enterprise data management.

**AI-Assisted Data Assessment and Data Extraction:** Concept of data profiling and data quality assessment. Techniques for evaluating completeness, uniqueness, and consistency of enterprise data. Introduction to extraction strategies including database extraction, API-based extraction, and file-based extraction. Use of AI-assisted tools to detect relationships, dependencies, and data patterns.

**Data Transformation and Data Quality Engineering:** Concept of source-to-target mapping in data migration. Data transformation techniques including data type conversion, code mapping, and lookup transformations. Data cleansing techniques such as duplicate detection and standardization. Introduction to data validation rules and quality scorecards.

**Data Loading, Reconciliation and Data Governance:** Data loading strategies in enterprise systems. Concepts of data reconciliation and validation after migration. Introduction to enterprise data governance including data cataloging, data lineage, stewardship, and compliance requirements. Overview of data quality monitoring and continuous data management.

## Reference Books

1. P. Jones, "The AI-Enabled Enterprise Data Management Bible", 1st Edition, Expert Data Publication, 2023.
2. Industry–Academia Joint Initiative, "AI-Powered Enterprise Data Management – Course Module", Enterprise Data Engineering Training Material.
3. R. Kimball and M. Ross, "The Data Warehouse Toolkit: The Definitive Guide to Dimensional Modeling", 3rd Edition, John Wiley & Sons, 2013.
4. W. H. Inmon, "Building the Data Warehouse", 4th Edition, John Wiley & Sons, 2005.
5. M. Kleppmann, "Designing Data-Intensive Applications: The Big Ideas Behind Reliable, Scalable, and Maintainable Systems", O'Reilly Media, 2017.

## Web Resources

1. Amazon Web Services – <https://aws.amazon.com/data>
2. Microsoft Data Documentation – <https://learn.microsoft.com/data>
3. Google Cloud Data Analytics – <https://cloud.google.com/data-analytics>
4. SAP Data Management – <https://www.sap.com/products/data-management.html>
5. ChainSys Data Management Products – <https://docs.chainsys.com/>

**Course Contents and Lecture Schedule**

<b>Module No.</b>	<b>Topic</b>	<b>No. of Periods</b>
1	<b>Enterprise Data Ecosystem &amp; ERP Fundamentals</b>	
	Introduction to Enterprise Resource Planning (ERP) systems	1
	Overview of major ERP platforms: SAP, Oracle, Salesforce, MS Dynamics	2
	Enterprise data types: Master data, Transactional data, and Metadata	1
	Enterprise data architecture and database structures	1
	Enterprise data lifecycle and migration overview	1
	Role of Artificial Intelligence in enterprise data management	1
2	<b>AI-Assisted Data Assessment and Extraction</b>	
	Fundamentals of data profiling and data assessment	1
	Data quality dimensions: completeness, consistency, and uniqueness	1
	Data extraction methods from enterprise systems	1
	Database-based data extraction techniques	2
	API and file-based data extraction methods	1
	Identification of data relationships and dependencies	1
3	<b>Data Transformation and Data Quality Engineering</b>	
	Source-to-target data mapping concepts	1
	Data transformation techniques in enterprise systems	1
	Data type conversion and code/value mapping	1
	Data cleansing and standardization methods	1
	Duplicate detection and data correction techniques	2
	Data validation rules and quality metrics	1
4	<b>Data Loading, Reconciliation and Data Governance</b>	
	Data loading strategies in enterprise systems	1
	Data migration and deployment approaches	1
	Post-migration reconciliation techniques	1
	Enterprise data governance principles	1
	Data catalog, data lineage, and stewardship concepts	2

	Data privacy, compliance, and security considerations	1
<b>Total</b>		<b>28</b>

**Course Designer(s):**

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**CURRICULUM FRAMEWORK AND SYLLABUS**  
**FOR**  
**FIVE YEAR INTEGRATED M.Sc. (DATA SCIENCE) DEGREE PROGRAMME**  
**IN CHOICE BASED CREDIT SYSTEM**  
**FOR THE STUDENTS ADMITTED FROM THE**  
**ACADEMIC YEAR 2023-2024 ONWARDS**  
**MOOC Courses**



**THIAGARAJAR COLLEGE OF ENGINEERING**  
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**23DSMA0****XAI: ESSENTIALS, CONCEPTS  
AND PRACTICE**Category L T P Credit  
PEC 1 0 0 1**Preamble**

This course provides an overview of Explainable AI (XAI) techniques, focusing on the importance of transparency and interpretability in machine learning models. It introduces interpretable models, model-agnostic methods, and, with real-world applications and ethical implications.

**Prerequisite**

- Nil

**Course Outcomes**

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

CO1	Explain the importance of interpretability in machine learning models and understand the trade-offs with model accuracy.	Understand
CO2	Describe interpretable models such as linear models and decision trees to provide transparent predictions.	Apply
CO3	Utilize local interpretability techniques, including LIME and SHAP, to provide explanations for individual predictions.	Apply
CO4	Assess ethical implications in AI systems, focusing on bias detection and fairness in interpretable models.	Analyze

**Mapping with Programme Outcomes**

Cos	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	L	M	M	L	L	L	L	L	L	L	S	M
CO2	S	S	M	M	L	L	L	L	L	L	L	L	S	M
CO3	S	S	M	S	M	L	L	L	L	L	L	L	S	S
CO4	S	M	S	M	M	M	M	S	L	L	L	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1 *						CAT2*						Terminal**					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																		
CO1	25	25											10	10				
CO2	10	20	20										5	10	15			
CO3							10	20	20				5	10	10			
CO4								10	20	20				5	10	10		
Total (in %)	100						100						100					

\* - WEEKLY QUIZZES

\*\*- PROCTORED QUIZ

**Syllabus**

**INTRODUCTION TO INTERPRETABLE MACHINE LEARNING-** Introduction to Interpretable Machine Learning, Why Interpretability Matters in Machine Learning, Black Box Vs White Box Models, Global Vs Local Interpretability, Trade off between Interpretability and Accuracy, Real Word Case Studies

**INTERPRETABLE MODELS-** Introduction to Linear models, Linear interpretability, Linear regression with categorical variables, Bias and Variance, Ridge Regression for improving the model, Lasso regression for improving the explainability, Feature subset selection

**LOCAL INTERPRETABILITY TECHNIQUES-** Local Surrogate Models (LIME), Shapley Values and SHAP explanations, Counterfactual Explanations, Anchors for decision rules, Case Study: Counterfactual explanations for loan approval decisions in banking systems

**ETHICAL IMPLICATIONS AND CHALLENGES-** Interpretability in healthcare, finance, and legal systems, Bias detection and fairness in AI, Ethical challenges in deploying interpretable models, Future directions in Explainable AI research, Case Study: Ethical implications of explainable AI in healthcare diagnosis models

#### Reference Books & web resources

1. Molnar, Christoph. Interpretable Machine Learning: A Guide for Making Black Box Models Explainable, 2022. (Chapters 1 to 9). <https://christophm.github.io/interpretable-ml-book/>
2. Explainable AI for Practitioners by Michael Munn and David Pitman, O'Reilly Media, 2022.
3. Thampi A. Interpretable AI: Building Explainable Machine Learning Systems. Leanpub; 2021.

#### Course Designers:

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**23DSMB0****PYTHON PROGRAMMING FOR  
DATA ANALYTICS**

Category	L	T	P	Credit
PEC	1	0	0	1

**Preamble**

This course introduces Python for Data Analytics and its applications in data-driven problem solving. It covers programming fundamentals, data manipulation using essential libraries, and techniques for data visualization and dashboard development. The course emphasizes practical implementation for analyzing and presenting insights from real-world datasets.

**Prerequisite**

- Basics of Programming Knowledge
- Basic mathematical skills
- Logical thinking ability

**Course Outcomes**

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

CO1	Understand basic Python programs using variables, data types, operators, loops, and data structures.	Understand
CO2	Use functions, NumPy, and Pandas to perform data manipulation and basic analysis.	Apply
CO3	Create simple data visualizations and basic interactive dashboards to present data clearly	Apply

**Mapping with Programme Outcomes**

Cos	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	L	M	M	L	L	L	L	L	L	L	S	M
CO2	S	S	M	S	S	L	L	L	L	L	L	L	S	S
CO3	S	M	S	M	S	M	L	L	M	L	L	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1 *						CAT2*						Terminal**					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																		
CO1	25	25											5	10				
CO2	10	20	20					10	20				5	10	15			
CO3							15	25	30				5	10	20			
Total (in %)	100						100						100					

\* - WEEKLY QUIZZES

\*\* - PROCTORED QUIZ

**Syllabus**

**Python Environment and Basics:** What is Python? – Introduction and Applications - Installing Anaconda and Launching Jupyter Notebook, Visual Studio Code – Jupyter and Visual Studio Code IDE overview - Creating, saving, and running script - Executing python in Google Colab - Python syntax and indentation - Writing comments in Python- Input and output in Python, Variables – Naming rules, Creation, Assignment - Data types, Operators and Decision Control Statements in Python.

**Sequences, Loops, and Basic Data Structures:** Strings – Lists - Tuples and Dictionaries - Sets and Range Function - Loops and Loop Control Statements - Experiments: Demonstration with Practice Exercises.

**Array, Functions, NumPy, and Pandas Basics:** Arrays – creation, indexing, slicing – array arithmetic, statistics, broadcasting – Functions: definition, arguments, return values – fruitful functions, recursion – list comprehension – Pandas basics: Series, DataFrames – reading CSV/Excel datasets – data exploration using head(), tail(), describe(), info() – filtering and selection.

**Data Visualization and Dashboards:** Introduction to data visualization - Matplotlib – scatter plot, line plot, bar plot, histogram, box plot - Introduction to dashboards – purpose and principles - Creating dashboards with Streamlit – simple interactive example

### Reference Books & web resources

1. Wes McKinney, “Python for Data Analysis: Data Wrangling with pandas, NumPy, and Jupyter”, 3rd Edition, O’Reilly Media, 2022. (Chapters 1–12).
2. Jake VanderPlas, “Python Data Science Handbook: Essential Tools for Working with Data”, O’Reilly Media, 2016. (Relevant chapters on NumPy, pandas, and Matplotlib).
3. Yuli Vasiliev, “Hands-On Data Analysis with Python: Perform data collection, wrangling, analysis, and visualization using Python”, 2nd Edition, No Starch Press, 2022.
4. Kaggle Learn: Hands-on exercises and datasets for practical data analytics learning. <https://www.kaggle.com/learn>
5. IBM SkillsBuild : Provides beginner-friendly learning paths in Python, AI, and Data Analytics. <https://skillsbuild.org>

### Course Designers:

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**23DSMC0****STATISTICS FOR MACHINE  
LEARNING**Category L T P Credit  
PEC 1 0 0 1**Preamble**

This course aims to introduce beginners to the fundamental concepts of statistics and probability that form the foundation of machine learning and data science. It develops learners' ability to understand, summarize, and interpret data using descriptive and inferential statistical techniques, probability distributions, and hypothesis testing methods. The course also emphasizes statistical thinking for problem solving and decision making, enabling students to apply appropriate statistical tools using Python/R for analyzing real-world datasets and supporting effective machine learning model development.

**Prerequisite**

- Nil

**Course Outcomes**

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

CO1	Understand the fundamental concepts of statistics and probability required for machine learning.	Understand
CO2	Develop the ability to describe and summarize data using statistical measures and visualizations.	Apply
CO3	Learn probability distributions and their relevance in modeling real-world data.	Apply
CO4	Apply inferential statistical techniques such as estimation and hypothesis testing for data-driven decision making.	Apply
CO5	Use Python/R tools to implement statistical methods and analyze practical machine learning datasets.	Apply
CO6	Build statistical thinking skills necessary for solving machine learning problems effectively.	Apply

**Mapping with Programme Outcomes**

Cos	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
CO1	S	M	L	M	M	L	L	L	L	L	L	L	S	M
CO2	S	S	M	M	L	L	L	L	L	L	L	L	S	M
CO3	S	S	M	S	M	L	L	L	L	L	L	L	S	S
CO4	S	M	S	M	M	M	M	S	L	L	L	L	S	S

S- Strong; M-Medium; L-Low

**Assessment Pattern**

CO	CAT1						CAT2						Terminal					
	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
Bloom's level																		
CO1	20	20											10	10				
CO2	10	10	20										10	10				
CO3			20												10			
CO4							20	20							20			
CO5							10	10	20						20			
CO6									20						10			
Total (in %)	100						100						100					

## Syllabus

### Introduction to Statistics for Machine Learning & Probability Fundamentals

Introduction – Relationship between statistics and data science – Role of statistical thinking in ML problem solving – Probability concepts and rules (addition, multiplication) – Conditional probability - Bayes' theorem – Random variables (discrete & continuous) – Expectation, Variance

### Probability Distributions & Descriptive Statistics

Binomial – Poisson – Uniform – Normal distribution – Descriptive Statistics: Measures of central tendency – Measures of dispersion - Visualization tools - Skewness, Kurtosis – Covariance – Correlation – Regression

### Inferential Statistics

Sampling and sampling distributions – Estimation – Confidence intervals – Hypothesis Testing: Null and alternative hypotheses – Type I & Type II errors – Critical values – t-tests – z-tests – Chi-square tests

### Practical Applications & Case Studies

Real-world data analysis projects – Use of Python/R to compute statistical methods – Visualization of statistical outputs

## Reference Books & web resources

1. James, G., Witten, D., Hastie, T., & Tibshirani, R. (2023). An Introduction to Statistical Learning: With Applications in Python (2nd ed.). Springer.
2. Wasserman, Larry. (2004). All of Statistics: A Concise Course in Statistical Inference. Springer.
3. Downey, Allen B. (2024). Think Stats: Exploratory Data Analysis in Python (3rd ed.). O'Reilly Media.
4. Stanford Statistical Learning Course – <https://online.stanford.edu/courses/soh-systatslearning-statistical-learning>
5. Scikit-learn User Guide – [https://scikit-learn.org/stable/user\\_guide.html](https://scikit-learn.org/stable/user_guide.html)

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