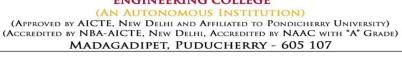


SRI MANAKULA VINAYAGAR ENGINEERING COLLEGE





B.Tech - Artificial Intelligence and Data Science M.Tech-Artificial Intelligence and Data Science

Minutes of 8th Board of Studies Meeting

Venue

AI & ML Lab, University Block, Lower Ground Floor Sri ManakulaVinayagar Engineering College Madagadipet, Puducherry - 605 107

> Date & Time 10.09.2024 & 10.30 AM Onwards

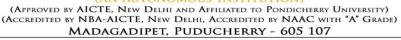




SRI MANAKULA VINAYAGAR

ENGINEERING COLLEGE

(AN AUTONOMOUS INSTITUTION)





The Eighth Board of Studies meeting for Department of Artificial Intelligence and Data Science was held on 10th September 2024 at 10:30 A.M in the Al & ML Lab, University Block, Lower Ground Floor, Sri Manakula Vinayagar Engineering College with the Head of the Department in the Chair.

The following members were present for the BoS meeting.

Sl. No.	Name of the Member	Designation
l. Head	of the Department concerned (Chairperson)	
1	Dr. J. Madhusudanan, M.E., Ph.D., Professor and Head Specialization: Ubiquitous and Edge Computing Years of Experience: 22 years Sri ManakulaVinayagar Engineering College hodaids@smvec.ac.in +91 90037 39274	Chairperson
2. All fac	culty members of the Department	
2	Dr. M.Auxilia. Associate professor Specialization: Cloud Computing, Deep Learning Years of Experience:19 years Sri Manakula Vinayagar Engineering College auxiliaaids@smvec.ac.in 9994276112	Member Secretary
3	Mr. K.Pragash, Assistant Professor, Specialization: Artificial Intelligence	Member
4	Mr. R.Rajan, Assistant Professor, Specialization: Machine Learning	Member
5	Mr.K.Muthukumaran, Assistant Professor Specialization: Cloud Security	Member
6	Mrs. M.Maragadhavalli Meenakshi, Assistant Professor, Specialization: Data Science, Deep Learning	Member
7	Ms.T,Shivaeeshwary, Assistant Professor, Specialization: Smart Computing	Member
8	Ms. S.Aishwarya Assistant Professor, Specialization: Machine Learning	Member
9	Mrs.S. Lakshmipriya, Assistant Professor, Specialization: Robotic Process Automation	Member
10	Mrs.P. Kanchanadevi, Assistant Professor, Specialization: Machine Learning, IoT	Member
11	Mrs.A.Ilakkiya Assistant Professor, Specialization: Smart Computing	Member
12	Mrs. V. Selvi, Assistant Professor Specialization: AI & ML	Member
13	Mrs.A. Keerthika, Assistant Professor Specialization: AI	Member



Mrs. N.Jayapratha, Assistant Professor Specialization: Networking	14 Spec 15 Mrs. Spec 16 Mrs. Spec 17 Mr. I Spec 18 Dr. N Spec 3. Two subject Dr. R Profe	Subashini M, Assistant Professor Subashini M, Assistant Professor Salization: Wireless Communication J. Roselin Lourd, Assistant Professor Salization: IoT Dhanapathy, Assistant Professor Salization: Computer Networks J. Ganesan, Professor / CSE Salization: Internet of Things experts from outside the Parent University are nominated b S. Srinivasa Perumal	Member Member Member Member				
Mrs. Subashini M, Assistant Professor Specialization:Wireless Communication	15 Mrs. Spec 16 Mrs. Spec 17 Mr. I Spec 18 Dr. M Spec 3. Two subject Dr. R Profe	Subashini M, Assistant Professor dalization:Wireless Communication J. Roselin Lourd, Assistant Professor dalization: IoT Dhanapathy, Assistant Professor dalization: Computer Networks J. Ganesan, Professor / CSE dalization: Internet of Things experts from outside the Parent University are nominated b J. Srinivasa Perumal	Member Member Member				
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Dr. R. Srinivasa Perumal Professor SCOPE 19a Vellore Institute of Technology, Vellore 8870537819 Mail id: Asstdean.acad3@vit.ac.in Dr. N. Bhalaji M.E., Ph.D Principal Rajalakshmi Institute of Technology 19b (An Autonomous Instituition) Chennai Ph:95000 86801 Mail id: bhalajin@ssn.edu.in 4. One expert is nominated by the Vice-Chancellor from a panel of six recommended by the Autonomous	Dr. R Profe	Srinivasa Perumal					
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College Principal as a University Nominee.	College Princi	oal as a University Nominee.					
Dr. N. Sreenath	Dr.	N. Sreenath					
Professor	Pro	essor					
Department of CSE	Der	artment of CSE					
20 Puducherry Technological University University University Nominee	20 Pud	ucherry Technological University	University Nominee				
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7. Experts from outside the Autonomous College, whenever special courses of studies are to be formulated				
nominated by the Principal.				
	Dr. V. Prasanna Venkatesan			
	Professor			
	Department of Banking Technology			
23	School of Management	Member		
	Pondicherry University			
	prasanna.btm@pondiuni.edu.in			
	+91 94887 34883			

AGENDA OF THE MEETING

B.Tech

- 1. Welcome Address and review of the seventh meeting of Board of Studies held on 04.03.2024.
- 2. To apprise about achievements of the institution and department
- 3. To discuss and approve syllabi of fifth and sixth Semesters for the B. Tech Artificial Intelligence and Engineering students admitted from the academic year 2023-24 under R-2023 Regulation.
- 4. To apprise and get approval for Honours Degree Courses, Syllabus and Credits
- 5. To discuss and get approval for evaluation Systems for regulation R-2023.
- 6. To discuss and get approval for the academic calendar for the odd semester 2024-25.
- 7. To apprise the End Semester Results of the students in the even semester
- 8. To apprise the schedule of the End Semester Examination to be conducted in the month of Nov/Dec 2024 and to discuss and recommend the panel of examiners to the Academic Council
- 9. To Discuss about Equivalence of Degree

M.Tech

- 1. To showcase the syllabus for the fourth semester M. Tech professional electives PE-IV, PE-V, PE-VI
- 2. To get suggestions from the members regarding the project domain selection and project proposal
- 3. To apprise about the List of Courses for Professional Electives / Ability Enhancement Courses / Mandatory Courses under R-2023 for the students admitted from the academic Year 2024-25.
- 4. To discuss about academic results

Minutes of the Meeting

Dr. J. Madhusudanan, Chairperson , BoS opened the meeting by welcoming and introducing the external members, to the internal and co-opted members and thanked them for accepting to become the member of the Board of Studies and the meeting thereafter deliberated on items that had been approved by the Chairperson .

BoS/8/2024/ AD /UG/8.1	The BoS Chairperson apprised the board regarding the minutes of 7 th BoS
BoS/8/2024/AD /UG/8.2	The BoS Chairperson apprised about the Achievements of College and Department. He portrayed the achievement of students in terms of co-curricular activities and placement records



BoS/8/2024/AD /UG/8.3	The BoS Chairperson apprised about the
	syllabus of V and VI Semester of R-2023. The
	syllabus was showcased to the BoS members
	and got concurrence and approval with minor
	modifications to be incorporated.
	1. The expert members have told to
	change the professional elective course
	Ethical Hacking U23ADE717 offered in
	VII semester to Quantum Al
	U23ADE717. The subject name has
	been changed.
	2. The subject code for the subject NLP
	and Chatbot has been changed from
	U23ADT611 to U23ADTC02 as it is
	going to be treated as common syllabus.
	AI & DS is going to offer this syllabus to
	CSE & BS. The expert members have
	ratified it.
	1

SI. No	Regulation	Semester	Subject Name with Code	Unit	Particulars
1	R-2023	V	Data Visualization U23ADT511	V	BoS Members have suggested to change the title from Introduction to Power BI to Power BI. The change has been incorporated. Kindly refer Annexure II
2	R-2023	VI	Web Technology U23ADT614	IV & V	BoS Members have suggested to change the contents of unit IV and V entirely to cover the broad aspects of PHP in web designing. Since CSE department is offering the same syllabus, with the concurrence of the BoS members department of CSE is offering syllabus for this subject under the name Web Designing / U23ADDC07. Kindly refer Annexure I and II
3	R-2023	VI	Data Science for Business Analytics U23ADE613	All units	The BoS Members have suggested to change the subject as it has replication of concepts in Machine Learning basics to Predictive Data Analytics / U23ADE612. The Subject has been changed. Refer Annexure I and II

BoS/8/2024/AD /UG/8.4	The BoS Chairperson apprised and got
	approval for Honours Degree – Courses,
	Syllabus and Credits. Refer Annexure III



BoS/8/2024/AD /UG/8.5	The BoS Chairperson discussed about evaluation Systems for regulation R-2023 and got approval for the same.
BoS/8/2024/AD /UG/8.6	The BoS Chairperson discussed about the academic calendar for the odd semester 2024-25 and got approval
BoS/8/2024/AD /UG/8.7	The BoS Chairperson apprised the End Semester Results of the students in the even semester. The BoS expert members appreciated the results and asked about the remedial measures for failed students and the BoS Chairperson explained the steps taken for remedial measures.
BoS/8/2024/AD /UG/8.8	The BoS Chairperson apprised the schedule of the End Semester Examination to be conducted in the month of Nov/Dec 2024 and showcased the panel of examiners.
BoS/8/2024/AD /UG/8.9	The BoS Chairperson discussed about Equivalence of Degree. He explained how the equivalence of 76% with Pondicherry University CSE curriculum and 80% with CSE curriculum of SMVEC. The BoS members have accepted the justification.

M.Tech Artificial Intelligence and Data Science

BoS/8/2024/ AD /PG/8.1	The BoS Chairperson apprised the board regarding the minutes of 7 th BoS The BoS Chairperson imparted about the list of subjects offered as professional electives PE-IV, PE-V, PE-VI and showcased the syllabus and the BoS members approved it gracefully. Kindly Refer Annexure IV and V
BoS/8/2024/AD /PG/8.2	The BoS Chairperson got suggestions from the members regarding the project domain selection and project proposal
BoS/8/2024/AD /PG/8.3	The BoS Chairperson apprised about the List of Courses for Professional Electives / Ability Enhancement Courses / Mandatory Courses under R-2023 for the students admitted from the academic Year 2024-25 and the BoS members approved it. Refer Annexure IV and V
BoS/8/2024/AD /PG/8.4	The BoS Chairperson showcased the results of the II semester and highlighted a student named Jeevitha. K has secured 10 on 10 CGPA in the semester exam and the BoS appreciated.

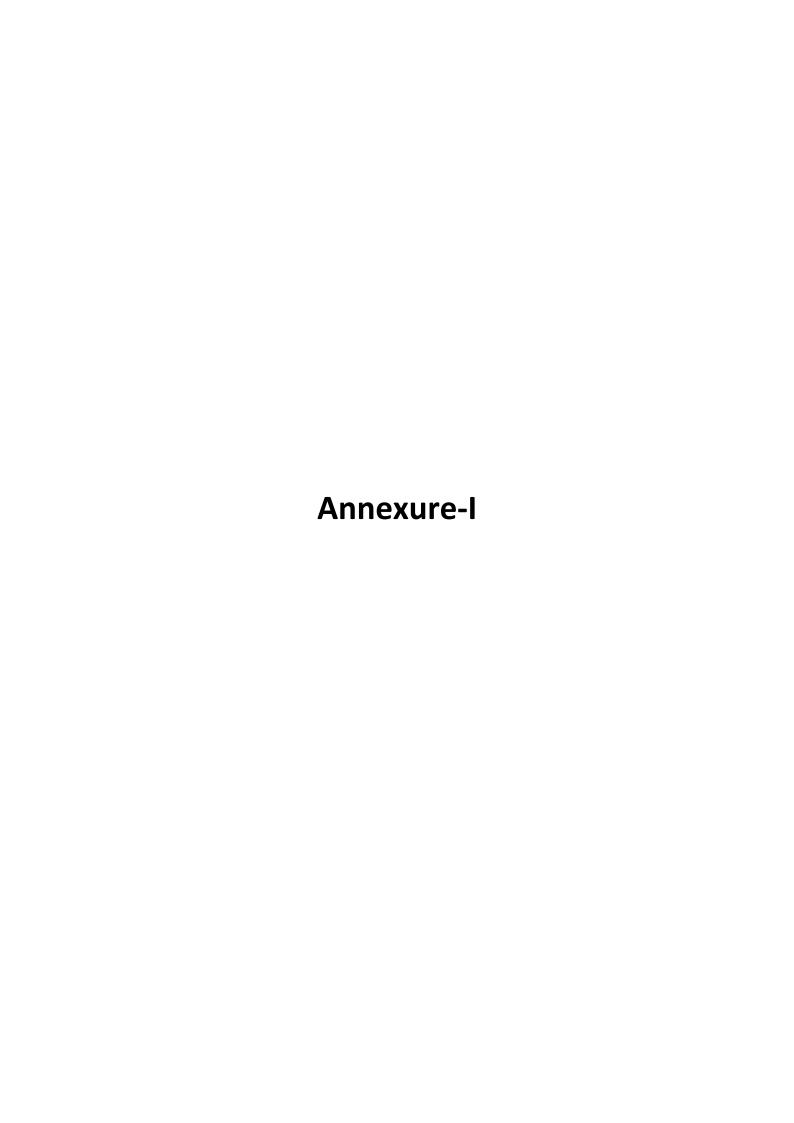


The meeting was concluded at 12.36 PM with a vote of thanks by **Dr. J. Madhusudanan**, Head of Department, Artificial Intelligence and Data Science.

Depa	rtment, Artificial Intelligence and Data Sc	dence.	
Sl. No	Name of the Member	Designation	Signature
1	Dr. J. Madhusudanan, M.E., Ph.D., Professor and Head Specialization: Ubiquitous and Edge Computing Years of Experience: 22 years Sri ManakulaVinayagar Engineering College hodaids@smvec.ac.in +91 90037 39274	Chairperson	5.1
2	Dr.N.Sreenath Professor Department of CSE Puducherry Technological University Puducherry Ph: 9443289642 Mail id: sreenath@ptuniv.edu.in	University Nominee	News
3	Dr.R.Srinivasa Perumal Professor SCOPE Vellore Institute of Technology, Chennai 8870537819 Mail id: r.srinivasaperumal@vit.ac.in	Subject Expert	R. Similas - Ruy -1
4	Dr. N. Bhalaji M.E., Ph.D Principal Rajalakshmi Institute of Technology (An Autonomous Instituition) Chennai Ph:95000 86801 Mail id: bhalajin@ssn.edu.in	Subject Expert	N. Post.
5	Dr. V. Prasanna Venkatesan Professor Department of Banking Technology School of Management Pondicherry University prasanna.btm@pondiuni.edu.in +91 94887 34883	Member	V. P V
6	Mr. E. Marie Joseph Antony Patrick Lead Software Engineer Freshworks Chennai Ph:9677488961 Mail id: patrick.ernest@freshworks.com	Industry Expert	Path

7	Ms. Madhu Srinvasan Engineer Director EMIS Health India Pvt. Ltd. Chennai Ph:99942 69567 Mail id: madhu_anusri@hotmail.com	Alumni	Madlin Ginevasan
8	Dr. M. Auxilia. Associate professor Specialization: Cloud Computing, Deep Learning Years of Experience: 19 years Sri Manakula Vinayagar Engineering College auxiliaaids@smvec.ac.in 9994276112	Member Secretary	Mhr
9	Mr. K.Pragash, Assistant Professor, Specialization:Artificial Intelligence	Member	8mit
10	Mr. R.Rajan, Assistant Professor, Specialization: Machine Learning	Member	Q. aexi
11	Mr.K.Muthukumaran, assistant Professor Specialization: Cloud Security	Member	Le Annot
12	Mrs. M.Maragadhavalli Meenakshi,Assistant Professor, Specialization: Data Science, Deep Learning	Member	DRYDRO
13	Ms.T,Shivaeeshwary, Assistant Professor, Specialization: Smart Computing	Member	1. Shivaling
14	Ms. S.Aishwarya Assistant Professor, Specialization: Machine Learning	Member	O. Ahuej
15	Mrs.S. Lakshmipriya, Assistant Professor, Specialization: Robotic Process Automation	Member	3 Lakestoni
16	Mrs.P. Kanchanadevi, Assistant Professor, Specialization: Machine Learning, IoT	Member	\$ 1
17	Mrs.A.Ilakkiya Assistant Professor, Specialization: Smart Computing	Member	Avida
18	Mrs. V. Selvi, Assistant Professor Specialization: Al & ML	Member	ish
19	Mrs.A. Keerthika, Assistant Professor Specialization:	Member	1.80
20	Mrs. N.Jayapratha, Assistant Professor Specialization: Networking	Member	NEG-

21	Mrs. Subashini M, Assistant Professor Specialization:Wireless Communicati on	Member	M. Suemains.
22	Mrs.J. Roselin Lourd, Assistant Professor Specialization: IoT	Member	5.2£
23	Mr. Dhanapathy, Assistant Professor Specialization: Computer Networks	Member	g. Lepothy
24	Dr. M. Ganesan, Professor / CSE Specialization: Internet of Things	Member	M. hare





(As per UGC - 2018 Regulations and Affiliated to Pondicherry University)

PUDUCHERRY - 605107

B.TECH. ARTIFICIAL INTELLIGENCE AND DATA SCIENCE Regulation-2023

CURRICULUM AND SYLLABI

COLLEGE VISION AND MISSION

Vision

To be globally recognized for excellence in quality education, innovation and research for the transformation of lives to serve the society.

Mission

M1: Quality Education:

To provide comprehensive academic system that amalgamates the cutting-edge technologies with best practices.

M2: Research and Innovation:

To foster value- based research and innovation in collaboration with industries and institutions globally for creating intellectuals with new avenues.

M3: Employability and Entrepreneurship:

To inculcate the employability and entrepreneurial skills through value and skill-based training.

M4: Ethical Values:

To instill deep sense of human values by blending societal righteousness with academic professionalism for the growth of society.

DEPARTMENT VISION AND MISSION

Vision

Incorporating the Data Science skills and applying the acquired analytical knowledge in the heterogeneous domains through Artificial Intelligence

Mission

M1: Understand Data Science:

Amalgamation of Programming Knowledge, Mathematical Skill Set and Knowledge of Business Domains to face the challenges of the real-world requirement

M2: Applying the Acquired Knowledge:

Inculcating the spirit of applying the acquired knowledge, innovation and creativity among students to work in heterogeneous domains

M3: Capstone Project:

Providing forum to carry out a capstone project through collaborations with the industries

5-11/-

M4: Be socially beneficial and other moral concerns:

Inspiring the educational experience in the field of application development and ensure the design, principle and ethic to be followed in the society.

M5: Continuous Learning for keen Initiative:

Affording continuous learning in the field of current trends in Artificial Intelligence and Data Science for keen initiative and enterprise focused.

PROGRAMME OUTCOMES (POs)

PO1: Engineering knowledge:

Apply the knowledge of mathematics, science, engineering fundamentals, and an engineering specialization to the solution of complex engineering problems.

PO2: Problem analysis:

Identify, formulate, research literature, and analyze complex engineering problems reaching substantiated conclusions using first principles of mathematics, natural sciences, and engineering sciences.

PO3: 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.

PO4: 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.

PO5: 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.

PO6: The engineer and society:

Apply reasoning informed by the contextual knowledge to assess societal, health, safety, legal and cultural issues and the consequent responsibilities relevant to the professional engineering practice.

5.15/-

PO7: 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.

PO8: Ethics:

Apply ethical principles and commit to professional ethics and responsibilities and norms of the engineering practice.

PO9: Individual and team work:

Function effectively as an individual, and as a member or leader in diverse teams, and in multidisciplinary settings.

PO10: 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.

PO11: 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.

PO12: Life-long learning:

Recognize the need for, and have the preparation and ability to engage in independent and lifelong learning in the broadest context of technological change.

PROGRAM EDUCATIONAL OBJECTIVES (PEOs)

PEO1: Acquiring the data:

To create an essential knowledge for extracting data from heterogeneous domains.

PEO2: Information Inferring and Knowledge representation:

To equip the student with knowledge, through different programming skills and creating a knowledge representation for the inferred data, so that it can be applied in the real time scenario.

PEO3: Design method:

To enable the student as a Data Analyst by designing a right Machine Learning algorithm and seamless programming skill to solve any sort of application.

PEO4: Systematic Enhancement:

To provide them with a keen knowledge on current trends and to enhance its impact periodically on the existing applications to meet the future scenario.

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PROGRAM SPECIFIC OBJECTIVES (PSOs)

PSO 1: Mathematical Foundation and Data Procuring:

To utilize the knowledge of Mathematical concept in procured Data from various Data sources.

PSO 2: Intellect Applications and Research Technologies:

To utilize the technical concepts, ideas, methodologies and the new emerging technologies in Artificial Intelligence and use this knowledge in their analytic skill to solving the current and future Data Analytics real time applications.

PSO 3: Developments of Real Time Applications:

To utilize the knowledge acquired and create a forum to carry out a capstone project through collaborations with the industries



STRUCTURE FOR UNDERGRADUATE ENGINEERING PROGRAM

SI. No	Course Category	Breakdown of Credits
1	Humanities and Social Sciences including Management courses (HS)	16
2	Basic Science Courses (BS)	17
3	Engineering Science Courses (ES)	41
4	Professional Core Courses (PC)	58
5	Professional Elective Courses (PE)	18
6	Open Elective Courses (OE)	09
7	Project Work and Internship (PA)	13
8	Ability Enhancement Courses (AEC*)	-
9	Mandatory Courses (MC*)	-
	Total	172

SCHEME OF CREDIT DISTRIBUTION - SUMMARY

SI.No	Course Category	Credits per Semester						per Semester To		
Oiiito	Godine Odlegory	I	II	III	IV	٧	VI	VII	VIII	Credits
1	Humanities and Social Science Courses (HS)	4	3	3	1	2	-	-	3	16
2	Basic Science Courses (BS)	4	4	5	4		-	-	-	17
3	Engineering Science Courses (ES)	12	12	6	11	-	-	-	-	41
4	Professional Core Courses (PC)	4	4	8	4	12	15	11	-	58
5	Professional Elective Courses (PE)	-	-	-	3	3	3	3	6	18
6	Open Elective Courses (OE)	-	-	-		3	3	3	-	09
7	Project Work (PA)	-	-	-	-	1	1	2	8	12
8	Internship (PA)	-	-	-	-	-	-	1	-	01
9	Ability Enhancement courses (AEC*) Courses (AEC*)	-	-	-	-	-	-	-	-	-
10	Mandatory Courses (MC*)	-	-	-	-	-	-	-	-	-
	* AEC and MC course Credits are not in			22	23	21	22	20	17	172

^{*} AEC and MC course Credits are not included for CGPA calculation

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HONOURS DEGREE PROGRAMME:

The student is permitted to opt for earning an honours degree in the same discipline of engineering in addition to the degree in his/her own discipline. To earn an honours degree the student is required to earn an additional 18 - 20 credits (over and above the total 170 credits prescribed in the curriculum) starting from fourth semester onwards by completing 5 additional courses offered in respective semesters. A student is eligible to exercise this option if he/she has passed all the courses offered upto third semester in the first attempt itself and has earned a CGPA / GPA* (*for lateral entry) of not less than 8.0. The prescribed courses offered for Honours degree are given in **Annexure V**.

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B. TECH CURRICULUM

	SEMESTER – I											
SI.	Course Code	Course Title	Category	P	Periods		'eriods		Credits	N	lax. Mar	ks
No.				L	Т	Р		CAM	ESM	Total		
Theory	Theory											
1	U23MATC01	Engineering Mathematics – I	BS	3	1	0	4	25	75	100		
2	U23ESTC03	Basics of Electrical and Electronics Engineering	ES	3	0	0	3	25	75	100		
3	U23CSTC01	Programming In C	ES	3	0	0	3	25	75	100		
4	U23ADT101	Digital System Design	ES	3	0	0	3	25	75	100		
5	U23ADT102	Fundamental of Data Science	PC	3	0	0	3	25	75	100		
Theory	Theory cum Practical											
6	U23ENBC01	Communicative English -I	HS	2	0	2	3	50	50	100		
Practic	al											
7	U23CSPC01	Programming in C Laboratory	ES	0	0	2	1	50	50	100		
8	U23ESPC03	Engineering Graphics using AutoCAD	ES	0	0	2	1	50	50	100		
9	U23ADP101	Fundamental of Data Science Laboratory	PC	0	0	2	1	50	50	100		
Ability Enhancement Courses												
10	U23ADC1XX	Certification Course-I**	AEC	0	0	4	0	100	-	100		
Mandatory Course												
11	11 U23ADM101 Induction Programme MC 2 Week			ks	0	-	-	-				
							22	425	575	1000		

	SEMESTER – II										
SI.	Course Code	Course Title	Category	Р	erio	ak	Credits	ľ	Max. Mark	(S	
No.	Course Code	Course Title	Category	L	Т	Р	Oreans	CAM	ESM	Total	
Theo	Theory										
1	U23MATC02	Engineering Mathematics – II	BS	3	1	0	4	25	75	100	
2	U23BSTC01	Physical Science for Engineers	ES	3	0	0	3	25	75	100	
3	U23ADTC01	Programming in Python	ES	3	0	0	3	25	75	100	
4	U23CSTC03	Data Structures	ES	3	0	0	3	25	75	100	
5	U23ADT203	Database Technologies	PC	3	0	0	3	25	75	100	
Theo	ry cum Practical									,	
6	U23ENBC02	Communicative English -II	HS	2	0	2	3	50	50	100	
Pract	ical						•				
7	U23ESPC02	Design Thinking and Idea Lab	ES	0	0	2	1	50	50	100	
8	U23ADPC01	Programming in Python Laboratory	ES	0	0	2	1	50	50	100	
9	U23CSPC02	Data Structures Laboratory	ES	0	0	2	1	50	50	100	
10	U23ADP202	Database Technologies Laboratory	PC	0	0	2	1	50	50	100	
Ability Enhancement Courses											
11	U23ADC2XX	Certification Course-II**	AEC	0	0	4	0	100	-	100	
Mand	Mandatory Course										
12	U23ADM202	Sports Yoga and NSS	MC	0	0	2	0	100	-	100	
							23	575	625	1200	

^{*} Certification Courses are to be selected from the list given in Annexure III

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	SEMESTER – III									
SI.	Course Code	Course Title	Category	Pe	rio	ds	Credits		Max. M	arks
No.	Godi So Godo	Source Title	Guiogory	L	T	Р	Orouno	CAM	ESM	Total
Theory										
1	U23MATC03	Probability and Statistics	BS	3	1	0	4	25	75	100
2	U23ADT304	Software Engineering and Agile Software Development	ES	3	0	0	3	25	75	100
3	U23ADT305	Artificial Intelligence and Expert System	PC	3	0	0	3	25	75	100
4	U23ADT306	Basic Machine Learning Techniques	PC	3	0	0	3	25	75	100
5	U23HSTC01	Universal Human Values-II	HS	2	0	0	2	25	75	100
Theo	Theory cum Practical									
6	U23CSBC01	Design and Analysis of Algorithms	PC	2	0	2	3	50	50	100
Pract	ical									
7	U23ENPC01	General Proficiency – I	HS	0	0	2	1	50	50	100
8	U23MAPC01	Engineering Mathematics Laboratory	BS	0	0	2	1	50	50	100
9	U23ADP303	Artificial Intelligence and Expert System Laboratory	PC	0	0	2	1	50	50	100
10	U23ADP304	Basic Machine Learning Techniques Laboratory	PC	0	0	2	1	50	50	100
Ability Enhancement Courses										
11	U23ADC3XX	Certification Course-III**	AEC	0	0	4	-	100	-	100
12	U23ADS301	Skill Enhancement Course-I*	AEC	0	0	2	-	100	-	100
Mandatory Course										
13	U23ADM303	Climate Change	MC	2	0	0	-	100	-	100
		·					22	675	625	1300

	SEMESTER – IV										
SI.	Course Code	Course Title	Cotomomi	Р	erio	ds	Cuadita		Max. Ma	rks	
No	Course Code	Course Title	Category	L	Т	Р	Credits	CAM	ESM	Total	
The	Theory										
1	U23MATC05	Discrete Mathematics and Graph Theory	BS	3	1	0	4	25	75	100	
2	U23ADDC01	Computer Networks and Security	ES	3	0	0	3	25	75	100	
3	U23ITTCO2	Programming in JAVA	ES	3	0	0	3	25	75	100	
4	U23ADT407	Advanced Machine Learning Techniques	PC	3	0	0	3	25	75	100	
5	U23ADE4XX	Professional Elective – I#	PE	3	0	0	3	25	75	100	
The	ory cum Practic	al	•								
6	U23ADB401	Linux Internals	ES	2	0	2	3	50	50	100	
Prac	ctical										
7	U23ENPC02	General Proficiency – II	HS	0	0	2	1	50	50	100	
8	U23ADP405	Computer Networks and Security Laboratory	ES	0	0	2	1	50	50	100	
9	U23ITPCO2	Programming in JAVA Laboratory	ES	0	0	2	1	50	50	100	
10	U23ADP406	Advanced Machine Learning Techniques Laboratory	PC	0	0	2	1	50	50	100	
Abil	ity Enhancemen	nt Courses									
11	U23ADC4XX	Certification Course-IV**	AEC	0	0	4	•	100	-	100	
12	U23ADS402	Skill Enhancement Course-II*	AEC	0	0	2	1	100	-	100	
Man	datory Course										
13	U23ADM404	Right to Information and Good Governance	MC	2	0	0	-	100	-	100	
							23	675	625	1300	

[#] Professional Elective Courses are to be selected from the list given in Annexure I *Skill Enhancement Courses (1 and 2) are to be selected from the list given in Annexure IV

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	SEMESTER – V										
SI.	Course Code	Course Title	Catamami	Pe	erio	ds	Credits	Max. Marks			
No.	Course Code	Course Title	Category	L	T	Р		CAM	ESM	Total	
Theory	Theory										
1	U23HSTC02	Research Methodology	HS	2	0	0	2	25	75	100	
2	U23ADT508	Cloud Computing and Data Management Architectures	PC	3	0	0	3	25	75	100	
3	U23ADT509	Deep Learning	PC	3	0	0	3	25	75	100	
4	U23ADT510	Data Visualization	PC	3	0	0	3	25	75	100	
5	U23ADE5XX	Professional Elective – II#	PE	3	0	0	3	25	75	100	
6	U23ADO5XX	Open Elective – I\$	OE	3	0	0	3	25	75	100	
Practic	Practical										
7	U23ADP507	Cloud Computing and Data Management Architectures Laboratory	PC	0	0	2	1	50	50	100	
8	U23ADP508	Deep Learning Laboratory	PC	0	0	2	1	50	50	100	
9	U23ADP509	Data Visualization Laboratory	PC	0	0	2	1	50	50	100	
Project	Work										
10	U23ADW501	Micro project	PA	0	0	2	1	100	-	100	
Ability	Ability Enhancement Courses										
11	U23ADC5XX	Certification Course-V**	AEC	0	0	4	-	100	-	100	
Manda	tory Course										
12	U23ADM505	Essence of Indian Traditional Knowledge	МС	2	0	0	-	100	-	100	
							21	600	600	1200	

SEMESTER – VI										
SI.	Course Code	SEWIES I	Category	F		ods	Credits		Мах. Ма	arks
No	00000	Course Title	Jacogory	L	. 1	Г	Orcuits	CAM	ESM	Total
Theory										
1	U23ADTC02	NLP and Chatbot	PC	3	0	0	3	25	75	100
2	U23ADT611	Robotic Process Automation – UI Path	PC	3	0	0	3	25	75	100
3	U23CSTC07	Web Designing	PC	3	0	0	3	25	75	100
4	U23ADE6XX	Professional Elective – III#	PE	3	0	0	3	25	75	100
5	U23ADO6XX	Open Elective – II \$	OE	3	0	0	3	25	75	100
Theory	cum Practical									
6	U23ADB602	Blockchain and Cryptography	PC	2	0	2	3	50	50	100
Practic	Practical									
7	U23ADP610	NLP and Chatbot Laboratory	PC	O	0	2	1	50	50	100
8	U23ADP611	Robotic Process Automation – UI Path Laboratory	PC	О) C) 2	1	50	50	100
9	U23CSPC06	Web Designing Laboratory	PC	О) (2	1	50	50	100
Project	t Work									•
10	U23ADW602	Mini project	PA	0) (2	1	100	-	100
Ability Enhancement Course										
11	U23ADC6XX	Certification Course – VI	AEC	0) () 4	-	100	-	100
Manda	tory Course									
12	U23ADM606	Gender Equality	MC	2	2 0	0	-	100	-	100
							22	625	575	1200

^{\$} Choose any one Open Elective Course from the list given in Annexure II

	SEMESTER - VII									
SI.	Course Code	Course Title	Catagory	P	Periods		Credits	Max. Marks		
No	Course Code	Course ride	Category	L	Т	Р	Credits	CAM	ESM	Total
Theo	Theory									
1	U23ADT712	Intelligent Systems and Control	PC	3	0	0	3	25	75	100
2	U23ADT713	IoT Systems and Analytics	PC	3	0	0	3	25	75	100
3	U23ADT714	Image Processing and Computer Vision	PC	3	0	0	3	25	75	100
4	U23ADE7XX	Professional Elective – IV#	PE	3	0	0	3	25	75	100
5	U23ADO7XX	Open Elective – III\$	OE	3	0	0	3	25	75	100
Prac	tical									
6	U23ADP712	Intelligent Systems and Control Laboratory	PC	0	0	2	1	50	50	100
7	U23ADP713	IoT Systems and Analytics Laboratory	PC	0	0	2	1	50	50	100
Proje	Project Work									
8	U23ADW703	Project Phase – I	PA	0	0	4	2	50	50	100
9	U23ADW704	Internship / In plant Training	PA	0	0	2	1	100	-	100
								375	525	900

	SEMESTER - VIII													
SI.				Р	Periods		Periods		Periods			Max. Marks		
No.	Course Code	Course Title	Category	L	Т	Р	Credits	CA M	ESM	Total				
Theo	Theory													
1	U23HSTC03	Entrepreneurship and Business Management	HS	3	0	0	3	25	75	100				
2	U23ADE8XX	Professional Elective – V#	PE	3	0	0	3	25	75	100				
3	U23ADE8XX	Professional Elective – VI#	PE	3	0	0	3	25	75	100				
Proje	Project Work													
4	U23ADW805	Project Phase – II	PA	0	0	16	8	50	100	150				
							17	125	325	450				

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ANNEXURE - I
PROFESSIONAL ELECTIVE COURSES (18 CREDITS)

SI. No.	Course Code	Course Title
	Professional Elec	tive – I (Offered in Semester IV)
1	U23CSDC01	Automata and Compiler Design
2	U23ADE401	Introduction to Computer Vision
3	U23ADE402	R Programming
4	U23ADE403	Tools and Techniques of Data Science
5	U23ADE404	Data Handling and Preprocessing
	Professional Electi	ive – II (Offered in Semester V) *
1	U23ADE505	Text Mining and Sentiment Analysis
2	U23ADE506	User Experience Design
3	U23ADE507	Java Programming: Essential Concepts to Advanced Mastery
4	U23ADE508	Exploratory Data Analysis
5	U23ADE509	Designing Machine Learning Systems
F	Professional Electi	ve – III (Offered in Semester VI) *
1	U23ADE610	Speech Processing and Analytics
2	U23ITEC05	Augmented Reality and Virtual Reality
3	U23ADE611	Advanced Java Programming
4	U23ADE612	Predictive Data Analytics
5	U23ADE613	Advanced Natural Language Processing
P	Professional Electiv	ve – IV (Offered in Semester VII) *
1	U23ADE714	Al Ethics
2	U23ADE715	Security in AI and ML
3	U23ADE716	Ethics in Data Science
4	U23ADE717	Cloud based Machine Learning Platforms
5	U23ADE718	Quantum AI
P	rofessional Electiv	ve – V (Offered in Semester VIII) *
1	U23ADE819	AI in Agriculture
2	U23ADE820	AI in Healthcare
3	U23ADE821	Stream Processing
4	U23ADE822	Sustainable Al
5	U23ADE823	AI in Finance
Р	rofessional Electiv	e – VI (Offered in Semester VIII) *
1	U23ADE824	Augmented Analytics
2	U23ADE825	Modern Cryptography
3	U23ADE826	AI in Automobile Industry
4	U23ADE827	Al in E-Commerce
5	U23ADE828	AI in Smart Cities



ANNEXURE - II

OPEN ELECTIVE COURSES (09 CREDITS)

S. No	Course Code	Course Title	Offering Department	Permitted Departments						
Open Elective – I / Open Elective-II (Offered in Semester V/VI) (Offered in Semester V for CSE, IT, MECH, Mechatronics, AI&DS) (Offered in Semester VI for EEE, ECE, ICE, CIVIL, BME, CCE)										
1	U23ADDC02	Principle of Artificial Intelligence and Machine Learning	AI&DS	EEE, ECE, CSE, IT, ICE, MECH, CIVIL, CCE, BME, Mechatronics						
2	U23ADOC01	Introduction to Data Science	AI&DS	EEE, ECE, CSE, IT, ICE, MECH, CIVIL, CCE, BME, Mechatronics						
Open	Elective - III (Offe	ered in Semester VII)								
3	U23ADOC02	Data science Application of Vision	AI&DS	EEE, ECE, CSE, IT, ICE, MECH, CIVIL, CCE						
4	U23ADOC03	Artificial Intelligence Applications	AI&DS	EEE, ECE, CSE, IT, ICE, MECH, CIVIL, CCE, BME, Mechatronics						

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ANNEXURE - III
ABILITY ENHANCEMENT COURSES-(A) CERTIFICATION COURSES

S. No	Course Code	Course Title	Certified By
1	U23XXCX01	Adobe Photoshop	Adobe
2	U23XXCX02	Adobe Animate	Adobe
3	U23XXCX03	Adobe Dreamweaver	Adobe
4	U23XXCX04	Adobe After Effects	Adobe
5	U23XXCX05	Adobe Illustrator	Adobe
6	U23XXCX06	Adobe InDesign	Adobe
7	U23XXCX07	Autodesk AutoCAD -ACU	Autodesk
8	U23XXCX08	Autodesk Inventor - ACU	Autodesk
9	U23XXCX09	Autodesk Revit - ACU	Autodesk
10	U23XXCX10	Autodesk Fusion 360 - ACU	Autodesk
11	U23XXCX11	Autodesk 3ds Max - ACU	Autodesk
12	U23XXCX12	Autodesk Maya - ACU	Autodesk
13	U23XXCX13	Cloud Security Foundations	AWS
14	U23XXCX14	Cloud Computing Architecture	AWS
15	U23XXCX15	Cloud Foundation	AWS
16	U23XXCX16	Cloud Practitioner	AWS
17	U23XXCX17	Cloud Solution Architect	AWS
18	U23XXCX18	Data Engineering	AWS
19	U23XXCX19	Machine Learning Foundation	AWS
20	U23XXCX20	Robotic Process Automation / Medical Robotics	Blue Prism
21	U23XXCX21	Advance Programming Using C	CISCO
22	U23XXCX22	Advance Programming Using C ++	CISCO
23	U23XXCX23	C Programming	CISCO
24	U23XXCX24	C++ Programming	CISCO
25	U23XXCX25	CCNP Enterprise: Advanced Routing	CISCO
26	U23XXCX26	CCNP Enterprise: Core Networking	CISCO
27	U23XXCX27	Cisco Certified Network Associate - Level 2	CISCO
28	U23XXCX28	Cisco Certified Network Associate- Level 1	CISCO
29	U23XXCX29	Cisco Certified Network Associate- Level 3	CISCO



30	U23XXCX30	Fundamentals Of Internet of Things	CISCO
31	U23XXCX31	Internet Of Things / Solar and Smart Energy System with IoT	CISCO
32	U23XXCX32	Java Script Programming	CISCO
33	U23XXCX33	NGD Linux Essentials	CISCO
34	U23XXCX34	NGD Linux I	CISCO
35	U23XXCX35	NGD Linux II	CISCO
36	U23XXCX36	Advance Java Programming	Ethnotech
37	U23XXCX37	Android Programming / Android Medical App Development	Ethnotech
38	U23XXCX38	Angular JS	Ethnotech
39	U23XXCX39	Catia	Ethnotech
40	U23XXCX40	Communication Skills for Business	Ethnotech
41	U23XXCX41	Coral Draw	Ethnotech
42	U23XXCX42	Data Science Using R	Ethnotech
43	U23XXCX43	Digital Marketing	Ethnotech
44	U23XXCX44	Embedded System Using C	Ethnotech
45	U23XXCX45	Embedded System with IOT / Arduino	Ethnotech
46	U23XXCX46	English For IT	Ethnotech
47	U23XXCX47	Plaxis	Ethnotech
48	U23XXCX48	Sketch Up	Ethnotech
49	U23XXCX49	Financial Planning, Banking and Investment Management	Ethnotech
50	U23XXCX50	Foundation Of Stock Market Investing	Ethnotech
51	U23XXCX51	Machine Learning / Machine Learning for Medical Diagnosis	Ethnotech
52	U23XXCX52	IOT Using Python	Ethnotech
53	U23XXCX53	Creo (Modelling & Simulation)	Ethnotech
54	U23XXCX54	Soft Skills, Verbal, Aptitude	Ethnotech
55	U23XXCX55	Software Testing	Ethnotech
56	U23XXCX56	MX-Road	Ethnotech
57	U23XXCX57	CLO 3D	Ethnotech
58	U23XXCX58	Solid works	Ethnotech
59	U23XXCX59	Staad Pro	Ethnotech
60	U23XXCX60	Total Station	Ethnotech
61	U23XXCX61	Hydraulic Automation	Festo
	1	1	i



63 U23XXCX63 Pneumatics Automation Festo 64 U23XXCX64 Agile Methodologies IBM 65 U23XXCX65 Block Chain IBM 66 U23XXCX66 Devops IBM 67 U23XXCX67 Artificial Intelligence ITS 68 U23XXCX68 Cloud Computing ITS 69 U23XXCX70 Cyber Security ITS 70 U23XXCX71 Data Analytics ITS 71 U23XXCX71 Data Analytics ITS 72 U23XXCX72 Databases ITS 73 U23XXCX73 Java Programming ITS 74 U23XXCX74 Networking ITS 75 U23XXCX75 Python Programming ITS 76 U23XXCX76 Web Application Development (HTML, CSS, JS) ITS 77 U23XXCX77 Network Security ITS & Palo alto 78 U23XXCX78 MATLAB Microsoft 80 U23XXCX78 Azure Fundamentals <th>62</th> <th>U23XXCX62</th> <th>Industrial Automation</th> <th>Festo</th>	62	U23XXCX62	Industrial Automation	Festo
65 U23XXCX65 Block Chain IBM 66 U23XXCX66 Devops IBM 67 U23XXCX68 Artificial Intelligence ITS 68 U23XXCX68 Cloud Computing ITS 69 U23XXCX70 Cyber Security ITS 70 U23XXCX71 Data Analytics ITS 71 U23XXCX72 Databases ITS 73 U23XXCX73 Java Programming ITS 74 U23XXCX73 Java Programming ITS 75 U23XXCX75 Python Programming ITS 76 U23XXCX76 Web Application Development (HTML, CSS, JS) ITS 77 U23XXCX76 Web Application Development (HTML, CSS, JS) ITS 78 U23XXCX77 Network Security ITS & Palo alto 79 U23XXCX78 MATLAB MathWorks 80 U23XXCX80 Azure Fundamentals Microsoft 81 U23XXCX81 Azure Data (DP-900) Microsoft 82 U23XXC	63	U23XXCX63	Pneumatics Automation	Festo
66 U23XXCX66 Devops IBM 67 U23XXCX67 Artificial Intelligence ITS 68 U23XXCX68 Cloud Computing ITS 69 U23XXCX69 Computational Thinking ITS 70 U23XXCX70 Cyber Security ITS 71 U23XXCX71 Data Analytics ITS 72 U23XXCX72 Databases ITS 73 U23XXCX73 Java Programming ITS 74 U23XXCX74 Networking ITS 75 U23XXCX75 Python Programming ITS 76 U23XXCX76 Web Application Development (HTML, CSS, JS) ITS 77 U23XXCX76 Web Application Development (HTML, CSS, JS) ITS 79 U23XXCX77 Network Security ITS & Palo alto 80 U23XXCX79 Azure Fundamentals Microsoft 80 U23XXCX80 Azure Fundamentals Microsoft 81 U23XXCX81 Azure Data (DP-900) Microsoft 82	64	U23XXCX64	Agile Methodologies	IBM
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85	83	U23XXCX83	Microsoft Security, Compliance and Identity (SC-900)	Microsoft
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92 U23XXCX92 Cyber Security Palo alto	90	U23XXCX90	Research Analyst	NISM
	91	U23XXCX91	Portfolio Management Services	NISM
93 U23XXCX93 Cloud Security Palo alto	92	U23XXCX92	Cyber Security	Palo alto
	93	U23XXCX93	Cloud Security	Palo alto



94	U23XXCX94	PMI – Ready	PMI
95	U23XXCX95	Tally – GST & TDS	Tally
96	U23XXCX96	Advance Tally	Tally
97	U23XXCX97	Associate Artist	Unity
98	U23XXCX98	Certified Unity Programming	Unity
99	U23XXCX99	VR Development	Unity

ANNEXURE - IV

ABILITY ENHANCEMENT COURSES-(B) SKILL ENHANCEMENT COURSES

SI. No.	Course Code	Course Title
	U23ADS301	SKILL ENHANCEMENT COURSE - I
1.	020/120001	a) Clean code
1.		b) Exploring of GITHUB
		c) Aptitude - I
	U23ADS402	SKILL ENHANCEMENT COURSE - II
2.		a) API design - I
		b) Exploring of Research Tools
		c) Aptitude - II

^{*} Choose any one SKILL ENHANCEMENT COURSE in the list for SEC - I, SEC - II

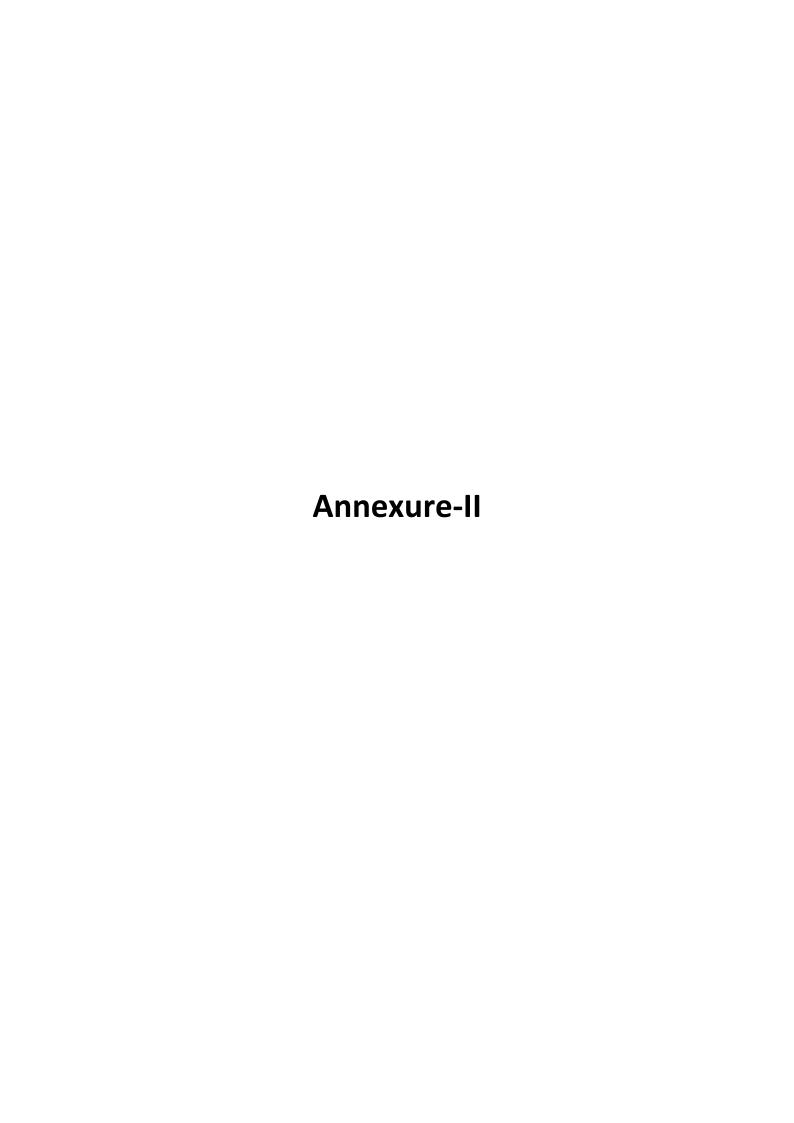
Annexure – V

HONOURS PROGRAMME – Artificial Intelligence and Machine Learning

COU	RSE DETAII	LS									
SI.	Semester	Category	Р	erio	ds	Credits	Ma	ax. Mark	s		
No.	Semester	Code	Course Title	Category	L	Т	Р	Credits	CAM	ESM	Total
Theo	ry			<u> </u>			l	1			
1	IV	U23ADX01	Parallel Programming and High Performance Computing	PC	3	1	0	4	25	75	100
2	V	U23ADX502	Advanced Deep Learning	PC	3	1	0	4	25	75	100
3	VI	U23ADX603	Reinforcement Learning	PC	3	1	0	4	25	75	100
4	VII	U23ADX704	Image and Video Analytics	PC	3	1	0	4	25	75	100
5	VIII	U23ADX805	Prompt Engineering	PC	3	1	0	4	25	75	100
	Total			<u> </u>				20	125	375	500
Equiv	valent NPTE	EL courses##			•			,			
1	Parallel C	Computing						3			
2	Deep Lea	arning for Con	nputer Vision					3			
3	Reinforce	ement Learnin		3	-	2 Weeks Course	5				
4	Image Pr	ocessing and (3							
5	Natural L	Language Proc	essing					3			

^{##} The student shall be given an option to earn 3 credits through one equivalent 12-week NPTEL course instead of any one course listed for honours degree programme that should be completed before the commencement of eighth semester. The equivalent courses are subject to change based on its availability as per NPTEL course list.

5-13/-



Department	Artificia	I Intelligence and Data Science	Programme	: B.Tech										
Semester	V		Course Cate	gory Code	e: HS *E	nd Sem	nester Ex	kam Typ	e: TE					
			F	Periods/W	eek	Credi	t Maximu	ım Mark	(S					
Course Code	Periods/Week Credit Maximum N													
Course	Researc	rch Methodology 2 2 25												
Name	NIII													
Prerequisite	INII							BT Ma	nnino					
	On com	pletion of the course, the students	s will be able t	to .				(Higl Lev	hest					
	CO1		arch methods	are used	to addre	ess eng	jineering	K	2					
		Students will develop the ab	oility to ident	ify resea	rch prob	olems,	perform	•						
Course	CO2	· ·	and use vario	us tools a	and service	ces for	effective	K	2					
Outcomes	CO3	Students will gain proficiency interpreting results using both num		•	-	zing da	ata, and	K	4					
		Students will be able to apply ethica				researc	h nanere							
	CO4	and dissertations, avoiding plagiari	•	oti dotaro e	and winter	Cocaro	η ραροίο	K	3					
		Students will understand the fund		tellectual	property	rights, i	ncluding							
						•								
	CO5	how to protect and enforce them, w	hich is crucial f	or innovat	ion and e	ntrepre	neurship	K	3					
		how to protect and enforce them, win engineering.	hich is crucial f	or innovat	ion and e	ntrepre	neurship	K	3					
UNIT-I	Introd	how to protect and enforce them, w in engineering. uction to Research				Pe	eriods:	6	3					
Meaning and of the Resea	Introd Importance	how to protect and enforce them, win engineering.	riew of Basic, Ap	plied, and E , Setting R	evelopme esearch C	Pental Res	eriods: (earch, Ov s and Re	6 rerview						
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Reference Books

- 1. Thiel, D. V. Research methods for engineers. Cambridge University Press 2014.
- 2. Ganesan, R. Research methodology for engineers. MJP Publishers 2024.
- 3. Agarwal, C & Sharma, V. Research methodology in sociology. Commonwealth Publishers 2012
- 4. Thody, A. Writing and presenting research (SAGE Study Skills Series). SAGE Publications 2006
- 5. Bordens, K. S. and Abbott, B. B, Research Design and Methods A Process Approach (d.) McGraw Hill, 8th Edition 2011.

Web References

- 1. https://conjointly.com/kb/
- 2. https://owl.purdue.edu/owl/research_and_citation/conducting_research/writing_a_literature_review.html
- 3. https://files.eric.ed.gov/fulltext/ED536788.pdf
- 4. https://researcheracademy.elsevier.com/
- 5. https://www.wipo.int/
- 6. https://www.scholastic.com/parents/school-success/homework-help/homework-project-tips/7-steps-to-successful-research-report.html
- 7. https://www.futurelearn.com/info/courses/business-research-methods- investigation.
- 8. https://articles.manupatra.com/article-details/Patent-Types-Laws-related-to-them-in-India
- 9. https://researchgate.net/
- 10. https://journals.sagepub.com/home/jmx

COs/POs Mapping

COs		Program Outcomes (POs)												ram Spe omes (P	
	PO1	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	3	2	2	1	1	1	1	1	1	1	3	2	2	2
2	2	3	2	2	2	1	1	1	2	2	1	3	2	2	1
3	3	3	3	3	2	1	1	1	1	1	2	2	2	2	1
4	2	2	1	2	1	1	1	3	2	3	1	2	2	3	1
5	2	2	2	2	1	2	2	3	2	2	3	3	2	2	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

		Conti	nuous Asse	ssment Marks (C	AM)	End Semester	
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Total Marks
Marks	1	0	5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

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Department	Artificial Intelligence and Data Science	Progran	nme: E	3.Tech.				
Semester	V	Course	Catego	ory Code	e: PC End	d Semeste	er Exam 1	¬уре: ТЕ
Course Code	U23ADT508	Perio	ds/We	ek	Credit	Ma	ximum M	arks
Course Code	023AD1300	L	Т	Р	С	CAM	ESE	TM
Course Name	CLOUD COMPUTING AND DATA	3	0	0	3	25	75	100
	MANAGEMENT ARCHITECTURES	3	U	U	3	25	/3	100
	F	N&DS					•	
Prerequisite	Database Technologies							
	On completion of the course, the stude	nts will b	e able	to			(⊢	Mapping lighest ₋evel)
	CO1 Interpret the concepts of Cloud Computi	ng and rea	lize it in	real wor	ld			K2
Course	CO2 Describe and realize the virtualization te	chniques t	o scale	up for re	sources			K2
Outcomes	CO3 Apply various cloud technologies and ac	lvancemer	ts for th	eir real-v	vorld scenario	S		K3
	CO4 Interpret various types of data available type of data	in the real	world a	nd apply	architecture b	ased on th	e K	(2, K3
	CO5 Apply the perfect data management arch	nitecture fo	r distrib	uted arch	nitecture and o	cloud		K3
Unit-I	Introduction to Cloud Computing				Periods: 9			
	oud Computing – Definition of Cloud – Evoluti			•		inciples of	Parallel a	
	uting – Cloud Characteristics – Elasticity in Cloud	l – On-den	nand Pro	ovisionin				CO1
UNIT-II	Cloud Enabling Technologies		<u>.</u>		Periods:9			
	Architecture – REST and Systems of Systems – V					• •		
-	Levels of Virtualization – Virtualization Structure lalization Support and Disaster Recovery	S – 100IS	and Med	cnanisms	s – virtualizati	on of CPU	-wemory	-
UNIT-III	Cloud Technologies and Advancement				Periods:9			
	uce-VirtualBox-Google App Engine-Programmir		nent for	Google /		nen Stack	-Federatio	าก
•	r Levels of Federation –Federated Services and	•		•		pon otaon		CO3
UNIT-IV	Data Management Architectures				Periods:9			L
Introduction to re	lational databases, -Database normalizations- L	imitations	of relat	ional dat	abases- Stru	ctured vs.	Unstructur	red CO4
data- Design of M	apReduce, Dataflow and Vertex-centric models	for proces	sing vol	ume, vel	ocity and linke	ed datasets	s-Storing a	ınd
querying over No	SQL datasets							
UNIT-V	Applications of Architecture for Manage	ement			Periods:9			
•	ms Architecture- Database Management Systen			•	•	g- Data Int	egration a	ind CO5
Processing Pipelii	nes- Data Indexing and Search- Data Visualization	on- Data S	ecurity a	and Priva	су			003
Lecture Perio	ds:45 Tutorial Periods:-	Practic	al Perio	ods:-	Т	otalPerio	ds:45	
Text Books					<u>.</u>			

- Buyya R., Broberg J., Goscinski A., "Cloud Computing: Principles and Paradigm", First Edition, John Wiley & Sons, 2011.
 Sosinsky B., "Cloud Computing Bible", 1st Edition, Wiley Edition, 2011
- 2. Michael Minelli, Michelle Chambers, and Ambiga Dhiraj, "Big Data, Big Analytics: Emerging Business Intelligence and Analytic Trends for Today's Businesses", Wiley, 1st Edition, 2013.
- 3. V.K. Jain, Big Data and Hadoop, Khanna Book Publishing Company 2020,2nd Edition.

Reference Books

- 1. Miller Michael, "Cloud Computing: Web Based Applications that Change the Way You Work and Collaborate Online", Pearson Education India, 1st Edition
- 2. Smooth S., Tan N., "Private Cloud Computing", Morgan Kauffman, 1st Edition, 2011.
- 3. EMC Education Services, "Data Science and Big Data Analytics: Discovering, Analyzing, Visualizing and Presenting Data", Wiley publishers, 1st Edition, 2015.
- 4. Bart Baesens, "Analytics in a Big Data World: The Essential Guide to Data Science and its Applications", 1st Edition, Wiley Publishers, 2015.

Academic Curriculum and Syllabi R-2023

Web References

- 1. https://onlinecourses.nptel.ac.in/noc21_cs14/preview
- 2. https://onlinecourses.nptel.ac.in/noc21_cs14/preview
- 3. https://archive.nptel.ac.in/courses/106/105/106105167/
- 4. https://archive.nptel.ac.in/courses/106/105/106105175/
- 5. https://onlinecourses.nptel.ac.in/noc23_ar01/preview

COs/POs/PSOs Mapping

		,a.	P9													
COs		Program Outcomes (POs)												Program Specific Outcomes (PSOs)		
	P01	PO2	PO3	PO4	PO5	PO6	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3	
1	2	3	2	2	2	-	-	-	-	-	-	-	2	3	3	
2	2	3	3	2	2	-	-	-	-	-	-	-	3	3	3	
3	2	2	3	2	2	-	-	-	-	-	-	-	2	3	3	
4	3	2	2	2	3	-	-	-	-	-	-	-	3	3	3	
5	2	3	2	2	3	-	-	-	-	-	-	-	3	3	3	

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Assessment		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

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Semester	Artificial Intelligence and Data Science	Programme	e: B.T	ech.					
Cemester	V	Course Ca	tegory	Code	: PC End	d Semeste	er Exam T	ype: Ti	
Course Code	1122 A DT500	Periods	/Week		Credit	Ma	ximum Ma	arks	
Course Code	U23ADT509	L	Т	Р	С	CAM	ESE	TM	
Course Name	DEEP LEARNING			İ					
		3	0	0	3	25	75	100	
	A	I&DS	.	<u>l</u> .			<u> </u>		
Prerequisite	Machine Learning								
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	On completion of the course, the studer	its will be a	bie to					ighest evel)	
	CO1 Interpret Neural Network basic Architecto	re and variou	ıs Activ	ation fu	unctions.		<u> </u>	K2	
Course									
Outcome								K2	
	CO3 Apply different optimization techniques to							K3	
	CO4 Interpret various deep learning models a correction		-		_	spelling		K2	
	CO5 Interpret various deep reinforcement tech	nniques for lea	arning	from fe	eback			K2	
UNIT-I	Foundations of Neural Networks				Periods: 9				
UNIT-II	rs: Learning Rate Momentum - Sparsity-Understar Convolutional Neural Network Blocks: Laver Type Convolutional Laver - Activa			α Lave	Periods:9	nnected La	aver -Bato	h co	
-	Blocks: Layer Type Convolutional Layer - Activa ropout Common architecture and Training Pattern	-			-		ayer -Batc	h CO2	
เพบเบเลแZสแบบ D	ropout Common architecture and Training Fattern	Leiver-2 - Ale	YING! A	9010	net - ResNet				
	•	Leivet-3 - Ale	XINEL V	0010		•			
UNIT-III	Regularization and Optimization				Periods:9		 		
UNIT-III Regularization D	Regularization and Optimization Propout Regularization Normalizing Inputs- Bootstr	ap Aggregatir	ng (Bag	ging)-	Periods:9 Dropout- Pro	s and Cons		al CO	
UNIT-III Regularization D Multitask Learni	Regularization and Optimization	ap Aggregatir nishing / Expl	ng (Bag oding (gging)- Gradier	Periods:9 Dropout- Pro nts - Weight	s and Cons	n Numerica		
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4. Josh Patterson, Adam Gibson, "Deep Learning A Practitioner's Approach", O'Reilly Media, 2017, Greyscale Indian Edition.

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Deng & Yu, "Deep Learning: Methods and Applications", Now Publishers, 2014
 Michael Nielsen, "Neural Networks and Deep Learning", Determination Press, 2015

5. Nikhil Buduma, "Fundamentals of Deep Learning", O'Reilly, 2017, first edition.

Academic Curriculum and Syllabi R-2023

Web References

- 1. https://nptel.ac.in/courses/106/106/106106184/
- 2. http://deeplearning.net/Dj
- 3. https://www.guru99.com/deep-learning-tutorial.html
- 4. https://www.coursera.org/specializations/deep-learning
- 5. http://neuralnetworksanddeeplearning.com/

COs/POs/PSOs Mapping

COs	Program Outcomes (POs)									Program Specific Outcomes (PSOs)					
	PO1	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	2	3	2	2	2	-	-	-	-	-	-	-	2	3	3
2	2	3	3	2	2	-	-	-	-	-	-	-	3	3	3
3	2	2	3	2	2	-	-	-	-	-	-	-	2	3	3
4	3	2	2	2	3	-	-	-	-	-	-	-	3	3	3
5	2	3	2	2	3	-	-	-	-	-	-	-	3	3	3

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Assessment		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

5.15/-

Department		ıal Intelli	gence and Data Science								
Semester	V			Course C	Categor	y Code:	PC E	nd Sen	neste	r Exam	Гуре: Т І
Course Code	U23AE	OT510		Perio	ds/We	ek	Credi	t	Maximum Marks		
	020/12			L	L T		С	CA	M	ESE	TM
Course Name	DATA	VISUALI	ZATION	3	0	0	3	2	5	75	100
				AI&DS				ii			
Prerequisite	Pythoi	n, Excel	and Data Science								
	On co	ompletio	n of the course, the stud	ents will b	e able	to			(BT Map Highest	
	CO1	Interpret	arious charts used and apply	y them accoi	rding to	the prob	em given			K2, K	(3
Course Outcome	CO2	Describe visualizat	various features used in seat on	oorn and Bol	keh and	apply th	em for data			K2, K	(3
Outcome	CO3	Interpret	now the data can be visualize	ed using Tab	leau					K2	
	CO4	Customiz	e and fine tune map aesthetic	cs using Tab	leau					K3	
	CO5	Apply Po	wer BI basics for generating i	nteractive re	ports ef	fectively				K3	
UNIT-I	Intro	duction t	- \/'!'				Periods	. ^	L		
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- 1. Daniel Nelson, "Data Visualization in Python", First Edition, StackAbuse, 2020.
- 2. Mamta Mittal, Nidhi Grover Raheja, "Data Visualization and Storytelling with Tableau", First Edition, CRC Press, 2024.
- 3. Jeremey Arnold, "Learning Microsoft Power BI", First Edition, O'Reilly Media, Inc, 2022.

Reference Books

- 1. Mario Dobler and Tim Grobmann, "Data Visualization with Python", Packt Publishing Ltd., 2019.
- 2. Praveen Kumar, "Data Visualization with Tableau", Gurucool Publishing, First Edition, 2020.
- 3. Seema Acharya., "Mastering Data Visualization using Tableau", First Edition, Wiley India Pvt.Ltd, 2024.
- 4. Chandraish Sinha, "Mastering Power BI", First Edition, BPB Publications, 2022
- 5. Greg Deckler, Brett Powell, "Microsoft Power BI Cookbook", Second Edition, Packt Publishing, 2021.

Web References

- 1. https://www.techtarget.com/searchbusinessanalytics/definition/data-visualization
- 2. https://machinelearningmastery.com/data-visualization-in-python-with-matplotlib-seaborn-and-bokeh/
- 3. https://www.tableau.com/learn/articles/data-visualization
- 4. https://help.tableau.com/current/pro/desktop/en-us/buildexamples_maps.htm
- 5. https://learn.microsoft.com/en-us/power-bi/fundamentals/power-bi-overview

COs/POs/PSOs Mapping

COs					Prog	ram O	utcom	es (PO	s)				Program Specific Outcomes (PSOs)		
	P01	PO1 PO2 PO3 PO4 PO5 PO6 PO7 PO8 PO9 PO10 PO11 PO1											PSO1	PSO2	PSO3
1	3	2	3	3	2	-	-	-	-	-	-	-	3	2	2
2	3	2	3	3	2	-	-	-	-	-	-	-	3	2	2
3	2	2	2	2	3	-	-	-	-	-	-	-	2	2	2
4	3	2	3	3	3	-	-	-	-	-	-	-	3	2	2
5	3	2	3	3	3	-	-	-	-	-	-	-	3	2	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Accomment		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	10		5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

5- NS/--

Department	Artific	cial Intelligence and Data Science	Program	nme: B.Te	ch.				
Semester	V		Course (Category	Code: I	PC	End Seme	ster Exar	n Type:
Course Code	1123 V	DP507	Perio	ds / Week	(Credit	Max	imum Ma	rks
Course Code	UZJA	DF 301	L	Т	Р	С	CAM	ESE	ΓМ
Course Name	MAN	JD COMPUTING AND DATA AGEMENT ARCHITECTURES DRATORY	0	0	2	1	50	50	100
		А	lands	<u> </u>		<u>i</u>	i	.i	<u>i</u>
Prerequisite	Databa	ase Technologies							
	On co	ompletion of the course, the studer	nts will be	able to					ipping t Level)
Course Outcomes	CO1	Configure various virtualization tools suc workstation	ch as Virtual	l Box, VMw		K 3			
	CO2	Simulate a cloud environment to implem	ent new sch	nedulers.				K	3
	CO3	Set up multi-node Hadoop Clusters.			K	3			
	CO4	Apply Map Reduce algorithms for various algorithms							3
	CO5	Apply instructed data processing using N	NoSQL and	data proce	ssing u	sing R pro	gramming	K	3

List of Experiments

- 1. Create a Collaborative learning environment for a particular learning topic using Google Apps. Drive, Google Docs and Google Slides must be used for hosting e-books, important articles and presentations respectively.
- 2. Install Virtual box and create a windows/linux virtual image and analyze the virtual configuration.
- 3. Install Google App Engine. Create hello world app and other simple web applications using python/java
- 4. Find a procedure to transfer the files from one virtual machine to another virtual machine.
- 5. Installation, Configuration, and Running of Hadoop and HDFS
- 6. Create a application for Page Rank Computation
- 7. Develop a MapReduce program to find the maximum temperature in each year
- 8. Develop a Java application to find the maximum temperature using Spark
- 9. Unstructured data into NoSQL data and do all operations such as NoSQL query with API.
- 10. Application to adjust the Number of Bins in the Histogram using R Language

Lecture Periods:	Tutorial Periods:	Practical Periods: 30	Total Periods: 30
Reference Books			

- EMC Education Services, "Data Science and Big Data Analytics: Discovering, Analyzing, Visualizing and Presenting Data", Wiley publishers, 1st Edition, 2015.
- 2. Bart Baesens, "Analytics in a Big Data World: The Essential Guide to Data Science and its Applications", Wiley Publishers, 2015.1st Edition.
- 3. Dietmar Jannach and Markus Zanker, "Recommender Systems: An Introduction", Cambridge University Press, 1st Edition, 2010.
- 4. Kim H. Pries and Robert Dunnigan, "Big Data Analytics: A Practical Guide for Managers" CRC Press, 2015,1st Edition.

Web References

- 1. https://davefoord.wordpress.com/2013/03/01/using-google-docs-drive-to-create-a-collaborative-learning-activity/
- 2. https://carleton.ca/scs/tech-support/virtual-machines/transferring-files-to-and-from-virtual-machines/
- 3. https://medium.com/@TadashiHomer/understanding-and-implementing-the-pagerank-algorithm-in-python-2ce8683f17a3
- 4. https://www.mongodb.com/resources/basics/databases/nosql-explained
- 5. https://www.naukri.com/code360/library/histogram-in-r-programming

COs/POs/PSOs Mapping

COs	Program Outcomes (POs)												Program Specific Outcomes (PSOs)			
	P01	PO1 PO2 PO3 PO4 PO5 PO6 PO7 PO8 PO9 PO10 PO11 PO													PSO3	
1	2	3	2	2	2	-	-	-	-	-	-	-	2	3	3	
2	2	3	2	2	2	-	-	-	-	-	-	-	3	3	3	
3	2	2	2	2	2	-	-	-	-	-	-	-	2	2	3	
4	3	2	2	2	3	-	-	-	-	-	-	-	3	3	3	
5	2	3	2	2	3	-	-	-	-	-	-	-	3	3	3	

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Assessment	Con	tinuous <i>A</i>	ssess	ment Marks (C/	AM)		
	Performano cla	e in pract sses	tical	Model		End Semester Examination	Total Marks
	Conduction of practical	Record work	viva	Practical Examination	Attendance	(ESE) Marks	Warks
Marks	15	5	5	15	10	50	100

Department	Artif	icial Intelligence and Data Science	Program	me: B. 1	Гесһ.				
Semester	٧		Course (ster Exa	am Type:				
			Perio	ods / W	eek	Credit	Max	imum M	arks
Course Code	U23A	DP508	L	Т	Р	С	CAM	ESE	TM
Course Name	DEEP	LEARNING LABORATORY	0	0	2	1	50	50	100
		AI&	DS	-					
Prerequisite	Machi	ne Learning							
Course	On co	empletion of the course, the students	will be ab	le to				(H	Mapping ighest .evel)
Outcomes	CO1	Apply neural network techniques to development networks for basic classification tasks.	velop and i	mpleme	nt simp	le feedfo	rward neu	ral	K 3
	CO2	Utilize convolutional neural networks (C classifying images	NN) to bui	ld deep	learnin	g model	s capable	of	K3

Implement recurrent neural networks (RNN) and LSTM models to predict time series data

Apply deep learning architectures like RNNs, LSTMs, and CNN for developing AI models for

Apply generative and transfer learning techniques to create artistic outputs

K3

K3

K3

List of Experiments

CO3

CO4

CO₅

- 1. Build a simple Neural Network.
- 2. Build a deep learning model to Classify cat and dog using CNN

domain-specific applications.

- 3. Build a deep learning model to predict Stock Prices using Recurrent Neural Network
- 4. Build a deep learning model to Forecast Sales using LSTM
- 5. Build a deep learning model to predict Movie box office using GRU model
- 6. Build a deep learning model to predict Sports result Prediction using RNN and LSTM
- 7. Build a deep learning model to predict cardiovascular disease using ANN
- 8. Build a deep learning model to create an art using Style Transfer technique
- 9. Build a deep learning model to a identify traffic signs from the image
- 10. Build a deep learning model for Fashion Recommendation System

Lecture Periods: Tutorial Periods: Practical Periods: 30 Total Periods: 30

Reference Books

- 1. Cosma Rohilla Shalizi, "Advanced Data Analysis from an Elementary Point of View", Cambridge University Press, 2015.
- 2. Deng & Yu, "Deep Learning: Methods and Applications", Now Publishers, 2014.
- 3. Michael Nielsen, "Neural Networks and Deep Learning", Determination Press, 2015.
- 4. Josh Patterson, Adam Gibson, "Deep Learning A Practitioner's Approach", O'Reilly Media, Greyscale Indian Edition, 2017.
- 5. Nikhil Buduma, "Fundamentals of Deep Learning", O'Reilly, 1st edition, 2017

Web References

- 1.https://nptel.ac.in/courses/106/106/106106184/
- 2. http://deeplearning.net/
- 3. https://www.guru99.com/deep-learning-tutorial.html
- 4. https://www.coursera.org/specializations/deep-learning
- 5. http://neuralnetworksanddeeplearning.com/

COs/POs/PSOs Mapping

COs		Program Outcomes (POs)													Program Specific Outcomes (PSOs)			
	PO1	PO1 PO2 PO3 PO4 PO5 PO6 PO7 PO8 PO9 PO10 PO11 PO1												PSO2	PSO3			
1	2	2 3 2 2 2							-	2	3	3						
2	2	3 3 2 2									-	3	3	3				
3	2	2	3	2	2	-	-	-	-	-	-	-	2	3	3			
4	3	2	2	2	3	-	-	-	-	-	-	-	3	3	3			
5	2	2 3 2 2 3												3	3			

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

	Con	itinuous <i>A</i>	ssess	ment Marks (C	AM)		
Assessment	Performano cla	e in pract	tical	Model		End Semester Examination	Total Marks
	Conduction of practical	Record work viva		Practical Examination	Attendance	(ESE) Marks	war KS
Marks	15	5	5	15	10	50	100

Department	Artificial Intelligence and Data Science Programme:B.Tech.										
Semester	V		Course	Catego	ry Cod		End Sem LE	Semester Exam Type:			
			Periods / Week			Credit	Ma	aximum Marks			
Course Code	U23AD	P509	L	Т	Р	С	CAM	ESE	TM		
Course Name	DATA	50	100								
	i	Al8	L								
Prerequisite	Pythor	and Excel									
	On co	npletion of the course, the students	will be abl	e to				BT Ma (Highest			
Course Outcomes	CO1	Apply data visualization techniques using graphs	Python libra	ries (Mat	plotlib,	Seaborn)	to create	K	3		
	CO2	Implement advanced visualization method and density plots for interpreting complex and tips.	•	•		•	•				
	CO3	Develop interactive visualizations using demographic, and market datasets	g Bokeh fo	r in-dep	th ana	alysis of	financial,	K	3		
	CO4	Apply Tableau to create complex data government budgets, and demographic st		ns for ar	nalyzinę	g stock m	narkets,	K	3		
	CO5	Apply Power BI dashboards to present ins geographic distributions.	ends, and	K 3							

List of Experiments

- 1. Create a bar graph and analyze on different social media platforms over the past month using Python.
- 2. Create a pie chart to split the equal halves of Demographic analysis using Python.
- 3. Create a line graph showing a diabetes using Seaborn.
- 4. Display a Dot Chart in Titanic dataset using Seaborn.
- 5. Create a Histogram and analyze a rainfall over month using Seaborn.
- 6. Create a Box Plots and analyze a Tips dataset using Bokeh.
- 7. Create a Density Plot and evaluate a Price distribution of Apartment in Airbnb Apartments using Bokeh.
- 8. Build a Scatter Plot and analyze a financial status of bank over the past month using Bokeh.
- 9. Create a Chart Tables in Government Budget using Tableau.
- 10. Create a Heat Map in Stock Market using Tableau.
- 11. Build a Population Pyramid in Tableau.
- 12. Create maps using Power BI.
- 13. Build a Web Analytics Dashboard using Power BI.

Lecture Periods:	Tutorial Periods:	Practical Periods: 30	Total Periods: 30

Reference Books

- 1. Mario Dobler and Tim Grobmann, "Data Visualization with Python", Packt Publishing Ltd., 2019.
- 2. Praveen Kumar, "Data Visualization with Tableau", Gurucool Publishing, 1st Edition, 2020.
- 3. Seema Acharya., "Mastering Data Visualization using Tableau", Wiley India Pvt.Ltd, 1st Edition, 2024.
- 4. Chandraish Sinha, "Mastering Power BI", BPB Publications, 1st Edition, 2022
- 5. Greg Deckler, Brett Powell, "Microsoft Power BI Cookbook", Packt Publishing, 2nd Edition, 2021.

Web References

- 1. https://www.techtarget.com/searchbusinessanalytics/definition/data-visualization
- 2. https://machinelearningmastery.com/data-visualization-in-python-with-matplotlib-seaborn-and-bokeh/
- 3. https://www.tableau.com/learn/articles/data-visualization
- 4. https://help.tableau.com/current/pro/desktop/en-us/buildexamples_maps.htm
- 5. https://learn.microsoft.com/en-us/power-bi/fundamentals/power-bi-overview

COs/POs/PSOs Mapping

COs					Prog	ram Oı	utcom	es (PO	s)				Program Specific Outcomes (PSOs)			
	PO1	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3	
1	3	2 3 3 2											3	2	2	
2	3	3 2 3 3 2										-	3	2	2	
3	2	2	2	2	3	-	-	-	-	-	-	-	2	2	2	
4	3	2	3	3	3	-	-	-	-	-	-	-	3	2	2	
5	3	3 2 3 3 3											3	2	2	

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

	Cor	ntinuous <i>A</i>	ssess	ment Marks (C	AM)		
Assessment	Performano cla	ce in pract	tical	Model		End Semester	Total Marks
	Conduction of practical	Record work	viva	Practical Examination	Attendance	Examination (ESE) Marks	Warks
Marks	15	5	5	15	10	50	100

Artificial Intelligence and Data Science

Department

Semester Course Code	V								
Course Code			······	Categor		e: PE *End	·····	r Exam Type:	TE
	U23A	DE505	Perio	ds/Wee	k	Credit	Ma	ximum Marks	
			L	Т	Р	С	CAM	ESE	TM
Course Name	TEXT I	MINING AND SENTIMENT YSIS	3	0	0	3	25	75	100
			AI&DS			•	•••••		
Prerequisite	Data S	Science, Natural Language Proce	ssing (NLP)						
	On co	mpletion of the course, the stu						BT Mapp (Highest L	
Course Outcomes	CO1	Understand and apply key text min unstructured text	ing technique	s for extra	acting u	useful informa	tion from	K2	
	CO2	Design sentiment analysis using be approaches	oth lexicon-ba	sed and ı	machin	e learning-bas	sed	K2	
	СОЗ	Implement machine learning mode	ls for text clas	sification	and se	entiment analy	sis	КЗ	
	CO4	identify various libraries and frame	works used in	text proc	essing	and sentimen	t analysis	K2	
	CO5	Apply advanced text mining techniq	ues to solve r	eal-world	proble	ms		КЗ	
UNIT – I	Introd	luction to Text Mining and Text	Preprocess	ing		Periods:9		-	
Text Preprocessi	ng - Toke gging, Na	g - Importance, challenges and Applicenization, stemming, and lemmatization amed Entity Recognition (NER) - Text	on - Stop word	l removal	, text n	ormalization a	nd case fol	ding - Part -of-	CO1
UNIT – II	Text F	Representation and Feature Ext	raction			Periods:9			i
ext Representat	tion Mode	els - Bag-of-Words (BoW) model: Cor	nstruction and	limitation	s - Ter	m Frequency	- Inverse D	ocument	T
Frequency (TF-IF)F) - Wor								
Embeddings - Se	-	d Embeddings - Introduction to Word mbeddings - N-grams (bigrams, trigrams)		nces bet	ween tr		/ TF- IDF a	and Word	CO2
Embeddings - Se PCA and LDA.	entence E	mbeddings - N-grams (bigrams, trigra	ams) and their	nces bet	ween tr	Dimensionality	/ TF- IDF a	and Word	CO2
Embeddings - Se PCA and LDA. UNIT – III	Text C	mbeddings - N-grams (bigrams, trigra	ams) and their	nces bet	ween tr nce - D	Dimensionality Periods:9	/ TF- IDF a	and Word echniques –	CO2
Embeddings - Se PCA and LDA. UNIT - III Supervised Learr Machines, Logisti	Text Ching for Textic Regres	mbeddings - N-grams (bigrams, trigra	ams) and their ing algorithms for IA) for topic m	nces beto	ween tr nce - D	Periods:9 on (Naïve Baye	reduction t	and Word echniques –	CO2
Embeddings - Se PCA and LDA. UNIT - III Supervised Learr Machines, Logisti	Text Coning for Text Coning for Text Con Control Contr	mbeddings - N-grams (bigrams, trigra Classification and Topic Modell ext Classification - Machine learning a sion) - Latent Dirichlet Allocation (LD	ams) and their ing algorithms for IA) for topic m	nces beto	ween tr nce - D	Periods:9 on (Naïve Baye	reduction t	and Word echniques –	
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Embeddings - Sepectal And LDA. UNIT - III Supervised Learn Machines, Logistic Learning for Text UNIT - IV Overview of Sentine Metrics UNIT - V Supervised Machapproaches to sepectal Andrew Service Lecture Perio 1. Matthew 2. Bing Liu	Text Coning for Text Coning fo	Classification and Topic Modell ext Classification - Machine learning (sion) - Latent Dirichlet Allocation (LD) ation - Handling Imbalanced Text Dark fluction to Sentiment Analysis (nalysis - applications, and challenges ons: SentiWordNet - Rule-based sentiment Analysis - Training (analysis). Aspect-Based Sentiment Analysis - Real-World Tutorial Periods: ell , "Mining the Social Web", O'Reilligen.	ing algorithms for the ing algorithms for topic matasets. in sentiment attiment analysist Analysis (ABSA) ld Applications Practically Media, Inc., 2017	text classodelling - analysis - s technique rning moderning modern	ween tr nce - E sification Identiff Types ues - S dels for ing ser ment A ds: -	Periods:9 on (Naïve Baye ying hidden to Periods:9 of sentiment entiment Anal Periods:9 resentiment an itiments towar nalysis To May 2012	reduction to reduction to reduction to reduction to reduction to reduction to reduction. Support to reduction	and Word echniques – Vector data - Deep eased Sentimen - Evaluation ep learning espects of a	CO3

2. Christopher D. Manning, Prabhakar Raghavan, and Hinrich Schütze "Introduction to Information Retrieval", Cambridge University

3. Julia Silge and David Robinson, "Text Mining with R: A Tidy Approach", O'Reilly Media, 2017

Programme: B.Tech.

5.15/-

Press, 2008.

Web References

- 1. https://towardsdatascience.com/sentiment-analysis-with-text-mining-13dd2b33de27
- 2. https://medium.com/@gladinv/introduction-to-text-mining-and-sentiment-analysis-affaaf520597

COs/POs/PSOs Mapping

COs					Prog	ram Oı	utcom	es (PO	s)					ram Spe omes (P	
COS	PO1	PO2	PO3	PO4	PO5	PO6	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	2	2	-	3	-	-	-	-	-	-	-	2	1	2
2	3	2	2	1	3	-	-	-	-	-	-	-	2	1	2
3	3	3	2	2	3	-	-	-	-	-	-	-	3	2	2
4	3	3	3	2	3	-	-	-	-	-	-	-	3	2	2
5	3	3	3	3	3	-	-	-	-	-	-	-	3	3	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Accomment		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	CAT 2	Attendance	Examination (ESE) Marks	Marks		
Marks	1	0	5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

2. 4 2/-

Department /	Artifici	al Intelligence and Data Science	Programi	ne: B.T	ech.				
Semester	V		Course C	ategory	/ Code:	PE *End	Semeste	r Exam Ty	ре: ТЕ
Course Code	1122 V	DE506	Perio	ds/Wee	ek	Credit	Ma	ximum Ma	rks
Course Code	UZSA	DE300	L	Т	Р	С	CAM	ESE	TM
Course Name	USER	EXPERIENCE DESIGN	3	0	0	3	25	75	100
			AI&DS					•	
Prerequisite	Softwa	are Engineering, Aesthetic sense							
•	On co	mpletion of the course, the stud	lents will be	e able t	0			BT Ma _l (Highest	
Course Outcomes	CO1	Interpret and apply the fundamental centric digital products.	principles of	user exp	perience	design to crea	ate user-	K2, F	(3
Cutcomes	CO2	Conduct effective user research, de inform design decisions.	velop user pe	ersonas,	and utiliz	ze usability tes	sting to	K3	
	СОЗ	Employ design thinking and strategy demonstrating the ability to develop	•			•		K3	
	CO4	Implement interaction design princip accessible and inclusive design practice.		arious de	evices an	d platforms, e	ensuring	K3	
	CO5	Evaluate user experience through appearing process for continuous impro					iterative	К3	
UNIT – I	Introd	luction to User Experience Desig	gn			Periods:9	<u></u>		
		mportance of user-centric design - Ov - Ethical Considerations in UX Design-l							t- CO
UNIT – II	Resea	arch in UX Design				Periods:9			
Research in Desig lata for design insi Research to Desig	ights-De	nniques for user research - Developing eveloping a Research Plan- Data Colle s and Resources.	g user person ection and Ar	as - Cor alysis-	nducting Synthesi	usability testir zing Researcl	ng - Analyz h Findings	ing researd Applying	h CO
UNIT – III	Desig	n Thinking and Strategy				Periods:9			
uccessful UX stra	tegies-	thinking - Frameworks for UX strate The Design Thinking Process, Design g- Ethical Considerations in Design T	Thinking To						on CO
UNIT – IV		ction Design	-			Periods:9			
lesign- Prototyping	and W	esign - Designing for different devices ireframing- Usability Testing, Responsi Design- Design Systems and Style.							CO
UNIT – V		raluation				Periods:9			
Methods for evalue	uating u	iser experience - Metrics and KPIs to Usability Testing, Analytics and Metric the Design Process- Case Studies and	cs- User Fee	dback ar	nd Suppo	cess - Imple			d co
		Tutorial Periods:-	Practic			To	tal Perioc	le:45	
Lecture Period	5.40	i utoriai rerious	Fractic	ai r tiil	Jus		iai i c iiot	13. 4 3	

- 1. Don Norman, "The Design of Everyday Things", Basic Books, 1st Edition, 2015
- 2. "Don't Make Me Think" Steve Krug, New Riders, 3rd Edition, 2014
- 3. Jeff Gothelf, Josh Seiden, "Lean UX: Designing Great Products with Agile Teams", Shroff/O'Reilly, 2nd Edition, 2016

Reference Books

- 1. Susan Weinschenk, "100 Things Every Designer Needs to Know About People", New Riders, 1st edition, 2011
- 2. "Measuring the User Experience: Collecting, Analyzing, and Presenting UX Metrics" Tom Tullis, Bill Albert. Morgan Kaufmann, 3rd Edition, 2022
- 3. Elizabeth Goodman, Mike Kuniavsky, Andrea Moed, "Observing the User Experience: A Practitioner's Guide to User Research", Morgan Kaufmann, 2nd Edition, 2012

Web References

- 1. https://www.uxdesigninstitute.com/blog/ux-design-principles/
- 2. https://imaginovation.net/blog/ux-in-product-development/
- 3. https://mailchimp.com/resources/how-to-create-a-user-persona-ux/
- 4. https://uxmag.com/articles/framework-for-designing-for-multiple-devices

COs/POs/PSOs Mapping

COs					Prog	ram O	utcom	es (PO	s)					ram Spe omes (P	
	P01	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	2	1	1	-	-	-	-	-	-	-	1	2	1	1
2	3	2	1	1	-	-	-	-	-	-	-	1	2	1	1
3	2	2	-	-	-	1	-	-	-	-	-	1	2	1	1
4	3	2	1	1	-	1	-	-	-	1	-	1	2	1	1
5	3	2	1	1	-	1	-	-	-	1	-	1	2	1	1

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

	Science	ntelligence and Data	Pro	ograr	nme	e: B.Tech	•		
Semester	V		Со	urse	Cat	tegory: P		End Semeste	r Exam Type: T
Course Code	U23ADE5	07		eriod: Wee	- 1	Credit		Maxim	num Marks
	020/1020		L	Т	Р	С	CAM	ESE	TM
Course Name	•	OGRAMMING: ESSENTIAL 'S TO ADVANCED	3	0	0	3	25	75	100
Prerequisite		erstanding of Java programment basics, database fundam		Con	сер	ts of obje	ct-oriente	d Programming,	, web
		etion of the course, the stud	dent	s wil					BT Mapping (Highest Leve
	CO1	Implementing Java collections							K3
	CO2	Interpret unit testing and mock							K2
Course Outcomes	CO3	Apply JDBC to interact with ar queries.						_	К3
Outcomes	CO4	Apply various techniques to cr sessions, and data presentation	on.					-	К3
	CO5	Implement AJAX to improve w data loading and interactions.	rep a	hhiics	uon	ı periormai	ice inrough	i asynchronous	К3
UNIT- I	Java Colle	ctions Framework				Periods:	9		
11117 11	Java Testi	ng Frameworks			Ī	Davis de	S		
functionalities - 、 Annotations : Co	JUnit: Overv JUnit with E mmon assert	ng Frameworks view and purpose of JUnit in clipse: Setting up and using methods and annotations in Jurkito for mocking objects in tests	JUnit nit - T	t in t	ne l	Eclipse ID e: Creating	9 eatures: K E - Assei	t Methods and	
Introduction to functionalities - A	JUnit: Overv JUnit with E mmon assert Mockito: Mock	riew and purpose of JUnit in clipse: Setting up and using methods and annotations in Jur	JUnit nit - T with	t in t Test S Real-	ne l	JUnit Fe Eclipse ID e: Creating	9 eatures: K E - Asser g and mana es.	t Methods and	
Introduction to functionalities - Annotations: Co Introduction to MUNIT- III Introduction to features and func (Data Manipulation	JUnit: Overv JUnit with E mmon assert Mockito: Mocl Database RDBMS: Bas tionalities of Con Language) ries and Proc	view and purpose of JUnit in clipse: Setting up and using methods and annotations in Jurkito for mocking objects in tests Management and SQL with ics of relational database management and square 11g - SQL Statements: S - DDL (Data Definition Language essing Results - Using Prepare	JUnit nit - T with I JDE agem elect e) - Ir	t in to rest Seal- BC nent seal- State ntrod	ne l uite Fim yste mer	Feriods: Periods: Periods: Priods: Priods: Priods:	9 eatures: K E - Asser g and mana es. 9 cle 11g In ting and So BC: Establis	troduction: Key bring Data - DML shing Connection	CO2
Introduction to functionalities - Annotations: Co Introduction to MUNIT- III Introduction to features and function to Introduction	JUnit: Overv JUnit with E mmon assert Mockito: Mock Database RDBMS: Bas tionalities of Con Language) ries and Procents and Trans	view and purpose of JUnit in clipse: Setting up and using methods and annotations in Jurkito for mocking objects in tests Management and SQL with ics of relational database management and square 11g - SQL Statements: S - DDL (Data Definition Language essing Results - Using Prepare	JUnit nit - T with I JDE agem elect e) - In ed St	t in to rest Seal- BC nent so State nert statem	ne l uite Fim yste mer	Feriods: Periods: Periods: Priods: Priods: Priods:	9 eatures: K E - Asser g and mana es. 9 cle 11g In ting and Sc BC: Establis Meta Data	troduction: Key bring Data - DML shing Connection	CO2
Introduction to functionalities - Annotations: Co Introduction to Introduction to Introduction to features and function (Data Manipulation Executing Quericallable Statement UNIT-IV Introduction to SET and POST re-Servlets Cookies	JUnit: Overv JUnit with E mmon assert Mockito: Mocl Database RDBMS: Bas tionalities of Con Language) ries and Procents and Trans Web Deve Servlets: Basi equests in ser es and Sessie	view and purpose of JUnit in clipse: Setting up and using methods and annotations in Jurkito for mocking objects in tests Management and SQL with ics of relational database management and sqL with pracle 11g - SQL Statements: S - DDL (Data Definition Languagessing Results - Using Preparactions.	JUniti - Time I JUNITI - Time	t in t t	uite Fim yyste mei ucti ents	Periods:	9 eatures: K E - Asser g and mana es. 9 cle 11g In ting and So BC: Establis Meta Data 9 d Post Rec context ha vlets - Intro	troduction: Key orting Data - DML shing Connection Objects - Using uests: Handling ndling in servlets oduction to JSP	CO2
ntroduction to functionalities - Annotations: Controduction to function to function to features and function to features and function function to function functio	JUnit: Overv JUnit with E mmon assert Mockito: Mocl Database RDBMS: Bas tionalities of Con Language) ries and Proc nts and Trans Web Deve Servlets: Basi equests in ser es and Sessi ges): Basics of Mastering	view and purpose of JUnit in clipse: Setting up and using methods and annotations in Jurkito for mocking objects in tests Management and SQL with ics of relational database management and square 11g - SQL Statements: S - DDL (Data Definition Language essing Results - Using Preparations. Iopment with Servlets and cs of servlet technology and life votes - Servlets Config and Coord Management: Managing coord JSP - JavaBeans in JSP: Using AJAX: Asynchronous Data	JUnit - Time in the result of	t in t	ne uite Fim yyste mer ucti sents	Periods:	9 eatures: K E - Asser g and mana es. 9 cle 11g In ting and Sc BC: Establis Meta Data 9 d Post Rec context ha evlets - Intro r encapsula	troduction: Key orting Data - DML shing Connection Objects - Using uests: Handling ndling in servlets oduction to JSP	CO2
ntroduction to functionalities - Annotations: Controduction to Multi- III ntroduction to eatures and function to Executing Querocallable Stateme UNIT- IV ntroduction to SET and POST reservicts Cookies Coo	JUnit: Overvice JUnit with Emmon assert Mockito:	view and purpose of JUnit in clipse: Setting up and using methods and annotations in Jurkito for mocking objects in tests Management and SQL with ics of relational database management and square 11g - SQL Statements: S - DDL (Data Definition Language essing Results - Using Preparations. Iopment with Servlets and cs of servlet technology and life votes - Servlets Config and Coord Management: Managing coord JSP - JavaBeans in JSP: Using AJAX: Asynchronous Data	JUniti - Time in the result of	t in to the time t	ne luite	Periods: ets Get and iration and sions in Seriods: es: Mechair	eatures: K E - Asser g and mana es. 9 cle 11g In ting and Sc C: Establis Meta Data 9 d Post Rec context ha volets - Intro r encapsula	t Methods and ging test suites - troduction: Key orting Data - DML shing Connection Objects - Using Indling in servlets oduction to JSP ting data.	CO2

Textbooks

- 1.Herbert Schildt, "Java: The Complete Reference",12th Edition, 2024.
- 2. Andy Hunt and Dave Thomas, "Pragmatic Unit Testing in Java 5.0: With JUnit", 2nd Edition,2022
- 3. Oracle Corporation, "Oracle Database 23c: New Features",2024
- 4. Bryan Basham, Kathy Sierra, and Bert Bates, "Head First Servlets and JSP: Passing the Sun Certified Web Component Developer Exam", 3rd Edition, 2023.

Reference Books

- 1. C.J. Date, "Database Management and SQL with JDBC", 2nd Edition, 2012.
- 2. Bryan Basham, Kathy Sierra, and Bert Bates, "Web Development with Servlets and JSP", 2nd Edition, 2008.
- 3. Joshua Bloch , "Effective Java", 3rd Edition, 2008.
- 4. David Flanagan, "JavaScript: The ,Definitive Guide", 2021.

Web References

- 1. https://archive.nptel.ac.in/courses/106/105/106105191/
- 2. https://www.tutorialspoint.com/java/index.htm
- 3. https://www.javatpoint.com/java-tutorial
- 4. https://www.geeksforgeeks.org/java/
- https://www.w3schools.com/java/

COs/POs/PSOs Mapping

COs			<u></u>		Pro	gram C	Outcom	es (PO	s)					gram Spe omes (P	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	3	2	2	3	2	-	-	-	-	-	-	2	2	-
2	2	2	2	3	3	2	-	-	-	-	-	-	3	1	-
3	3	3	2	2	2	2	ı	ı	-	-	-	ı	2	2	1
4	2	2	3	2	3	3	ı	ı	-	-	-	-	2	3	-
5	2	2	3	2	3	3			-	-	-	-	2	3	-

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Accessment		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	CAT 2	Attendance	Examination (ESE) Marks	Marks		
Marks	1	0	5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

Department	Artifi	cial Intel	ligence and Data Science	Progran	nme: E	3.Tech.				
Semester	V			Course	Catego	ory Cod	e: PC En	d Semeste	er Exam T	ype: TE
Course Code	11234	DE508		Perio	ods/We	ek	Credit	Ma	ximum Ma	arks
Course code	0237	DESCO		L	Т	Р	С	CAM	ESE	TM
Course Name	EXPL	ORATO	RY DATA ANALYSIS	3	0	0	3	25	75	100
			А	il&DS						
Prerequisite	Basic	Statistics	, Data Visualization and Pro	grammin	g in Pyt	thon				
	On co	ompletio	n of the course, the stude	nts will b	e able	to			(H	Mapping ighest evel)
0	CO1	Interpret	the need of exploratory data an	alysis						K2
Course Outcome	CO2	Identify th	ne usage of various python libra	ries and fu	unctions	for ED/	4			K2
34.001110	CO3	Apply un	variate data exploration and an	alysis for I	ΞDA					K3
	CO4	Apply biv	ariate data exploration and ana	lysis for E	DA					K3
	CO5	Identify v	arious techniques needed for ti	me series	analysis	3				K2
UNIT-I	Intro	duction 1	o Exploratory Data Analys	sis			Periods:	9	<u>-</u>	
•	lexing -	- Combinii	Pandas Objects – Data Indexing datasets – Concat, Append,	-		-	-	-	-	
UNIT-III	Univa	ariate An	alysis				Periods:9			
Tendency – Meas	sures of	Spread -	stribution Variables – Numeri Shape of the Distribution – Dat sis for Categorical Data.				•			:
UNIT-IV	Bivar	iate Ana	lysis				Periods:9			
			bles – Percentage Tables – Bivariate Analysis Methods –							
UNIT-V	Multi	variate a	nd Time Series Analysis				Periods:9			
			sal Explanations – Three-Varia - Data Cleaning – Time based i						tals of TSA	` _ CO5
Lecture Period	ds:45		Tutorial Periods:-	Practic	al Perio	ods:-	-	TotalPerio	ds:45	
Text Books				4			<u>L</u>			
 Wes McKinne Jake Vander 	ey, "Pyth Plas, "P	non for Da Python Dat	man Ahmed, "Hands-On Explo ta Analysis", Second Edition, P a Science Handbook: Essentia S. Stoffer, "Time Series Analy	ublished b I Tools for	y O'Reil Working	ly Media g with D	a, 2017. ata", Second	Edition, O F	Reilly, 2023	

- Reference Books
- 1. Eric Pimpler, "Data Visualization and Exploration with R", second edition, GeoSpatial Training service, 2020.
- 2. Claus O. Wilke, "Fundamentals of Data Visualization", second edition, O'reilly publications, 2023.
- 3. Matthew O. Ward, Georges Grinstein, Daniel Keim, "Interactive Data Visualization: Foundations, Techniques, and Applications", third Edition, CRC press, 2023.
- 4. Richard A. Johnson and Dean W. Wichern, "B Applied Multivariate Statistical Analysis", 7th Edition, Pearson, 2022.

Web References

- 1. https://www.geeksforgeeks.org/exploratory-data-analysis-in-python/
- 2. https://towardsdatascience.com/a-gentle-introduction-to-exploratory-data-analysis/
- 3. https://www.analyticsvidhya.com/blog/2021/08/exploratory-data-analysis-and-visualization-techniques-in-data-science/
- 4. https://www.coursera.org/learn/exploratory-data-analysis/

COs/POs/PSOs Mapping

COs		Program Outcomes (POs)												ram Spe omes (P	
	P01	PO2	PO3	PO4	PO5	PO6	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	2	2	3	3	3	2	-	-	-	-	-	-	2	2	2
2	2	2	3	3	2	2	-	-	-	-	-	-	2	2	2
3	2	2	2	2	1	2	-	-	-	-	-	-	2	3	1
4	2	3	2	2	2	1	-	-	-	-	-	-	2	3	1
5	3	3	3	3	3	2	-	-	-	-	-	-	2	2	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

A 64	cocomont		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
AS	sessment	CAT 1	CAT 2	Model Exam	Assignment*	Seme Examir (ESE) M	Examination (ESE) Marks	Marks
	Marks	1	0	5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

Department	Artificial Intelli	gence and Data Science	Progran						
Semester	V		Course	Catego	ry Code	e: PE End	d Semeste	er Exam 🛚	Гуре: Т
Course Code	U23ADE509		Perio	ds/We	ek	Credit	Ma	ximum M	arks
			L	Т	Р	С	CAM	ESE	TM
Course Name	DESIGNING M SYSTEMS	ACHINE LEARNING	3	0	0	3	25	75	100
			Al&DS				<u>.</u>		
Prerequisite	Basic understar	nding of machine learning o	concepts a	nd Prof	iciency i	n Python pro	ogrammin	g	
	On completio	n of the course, the stud	ents will b	e able	to			(⊢	Mappin lighest ₋evel)
Carras	CO1 Interpret	the fundamental components	and design	principle	es of mad	chine learning	systems.		K2
Course Outcome	CO2 Interpret	effective data preprocessing a	and feature e	engineei	ring techr	niques.			K2
	CO3 Deploy m	nachine learning models to pro	oduction env	vironmer	nts and m	onitor system	n performa	nce.	K3
	CO4 Deploy m	nachine learning models with a	appropriate (evaluatio	on metric	S.			K3
	CO5 Optimize	, maintain, and scale machine	e learning sy	stems, c	ulminatir	ng in a capsto	ne project.		K3
Preprocessing fo Engineering and normalization, and	Data types, source r ML Systems: D transformation: F d encoding -Data	-	solutions: re	A, t-SNE	E), Data	augmentation cal vs. nume	n techniqu	es - Feati	ıre
UNIT-III	Model Building	and Iraining ms: Overview of different M	l modolo /r		ا مامماما	Periods:9	rol potworl	(a) ahaaa	in al
he right mode I- idvanced tuning .2)- Early stoppin	Model Training techniques-Crossig, dropout, and da	and Hyperparameter Tunin validation techniques- Handliata augmentation	ng: Training	process	overvie	w-Grid searc tting: Regula	h, random	search, a	nd CO
UNIT-IV		tion and Deployment	Validation	and Ta	otina. Tr	Periods:9	V fold oro	oo volidati	
Model generaliza	tion and robustnes	ssion & classification- Model ss- Model Deployment Pipel ensorFlow Serving							
UNIT-V		otimization, and Scaling				Periods:9			
	re)- Optimization	ing Machine Learning Syste Techniques: Performance o							
Lecture Perio	ds:45	Tutorial Periods:-	Practic	al Perio	ods:-	T	otalPerio	ds:45	
Lecture Feilo									

- Validappa Lakshmanan, Sara Robinson, Michael Muhn, Machine Learning Design Patterns, First Edition, O'Reilly Media, 2020.
 Aurélien Géron, "Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow," Second Edition, O'Reilly Media, 2019.

Reference Books

- 1. Emmanuel Ameisen, "Building Machine Learning Powered Applications: Going from Idea to Product," First Edition, O'Reilly Media, 2020.
- 2. Giuseppe Bonaccorso, "Machine Learning Algorithms," Second Edition, Packt Publishing, 2020.
- 3. Ethem Alpaydin, "Introduction to Machine Learning," Fourth Edition, MIT Press, 2020.

Web References

- 1. https://github.com/chiphuyen/dmls-book
- 2. https://towardsdatascience.com/data-pipeline-design-patterns-100afa4b93e3
- 3. https://github.com/tensorflow/tensorflow

COs/POs/PSOs Mapping

COs					Prog	ram O	utcom	es (PO	s)					ram Spe omes (P	
	P01	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	3	3	3	2	-	-	-	-	-	-	2	1	3	2
2	3	3	3	2	2	-	-	-	-	-	-	2	1	3	2
3	3	3	3	2	2	1	-	-	-	-	-	3	1	3	2
4	3	3	3	2	2	-	-	-	-	-	-	3	1	3	2
5	3	3	3	2	2	-	-	-	-	-	•	3	1	3	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Accessment	CAT 1 CAT 2		s Assessmei	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	10		5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

Department	Artific	ial Intelligence and Data Science	Prog	ramme:	B. Tec	h.			
Semester	V		Cour	se Cate	gory Co	de: PA	*End Se	emeste	Exam Type: -
Course	1122 41	DW501	Pe	riods / V	Veek	Credit		Maxim	um Marks
Code	UZSAI	JW301	L	Т	Р	С	CAM	ESE	TM
Course Name	MICR	O PROJECT	0	0	2	1	100	-	100
		Al	& DS		<u>-</u>				
Prerequisite	Artifi	cial Intelligence, Machine Learning, D	eep L	earning,	Progra	mming in	C, Pytho	n, Java	l
	On co	mpletion of the course, the studen	ts will	be able	to				BT Mapping (Highest Level)
Course	CO1	Identify the problem statement for t survey	he mic	ro proje	ct work	through	the literat	ture	K2
Outcomes	CO2	Choose the proper components system.	as pe	r the r	equiren	nents of	the des	ign/	K2
	СОЗ	Apply the acquainted skills to devel	op fina	ıl model/	system/				К3

There shall be a Micro Project, which the student shall pursue as a team consists of maximum 4 students during the third year, fifth semester. The aim of the micro project is that the student has to understand the real time hardware / software applications. The student should gain a thorough knowledge in the problem he/she has selected and in the hardware / software he/she using in the Project. The Micro-project is an application that should be formally initiated and should be developed and also to be implemented by the respective team.

The Micro Project shall be submitted in a report form along with the hardware model / software developed, duly approved by the department internal evaluation committee. It shall be evaluated for 100 marks as Continuous Assessment. The department internal evaluation committee shall consist of faculty coordinator, supervisor of the project and a senior faculty member of the department. There shall be two reviews that will be considered for assessing a Micro Project work with weightage as indicated evaluation Methods.

Lecture Ferious Fractical Ferious. 30 Total Ferious. 30	Lecture Periods: -	Tutorial Periods: -	Practical Periods: 30	Total Periods: 30
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COs/POs/PSOs Mapping

, O, . O	U, . U U	ין קאיייי ט	J9												
COs					Prog	ram O	utcom	es (PO	s)				Program Specific Outcomes (PSOs)		
	PO1	PO2	PO3	PO4	PO5	PO6	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	2	2	2	-	-	-	-	3	3	-	1	1	1	1
2	3	3	3	2	2	2	2	2	3	3	3	1	2	2	2
3	3	2	2	1	-	2	-	-	3	3	3	1	3	3	3

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

		Review 1			Review 2			Total
Assessment	Novelty	Presentation	Viva	Presentation	Demonstration	Viva	Report	Marks
Marks	10	20	10	20	20	10	10	100

Department	Artificial Intelligence and Data Science	Prog	ramme:	B. Tech.				
Semester	V	Cours	e Catego	ry: AEC	End Se	mester E	Exam Ty	pe: -
0	HOLADOLYA	P	eriods/W	eek/	Credit	Max	imum Ma	arks
Course Code	U23ADC5XX	L	Т	Р	С	CAM	ESE	TM
Course Name	CERTIFICATION COURSE - V	0	0	4	-	100	-	100
	·				-			
Prerequisite	-							

Students shall choose an International / Reputed organization certification course of 40-50 hours duration specified in the curriculum (It is mandatory to do a minimum of six courses) which will be offered through the Centre of Excellence. These courses have no credit and will not be considered for CGPA calculation.

- (i) Certification Courses are required to be completed to fulfil the degree requirements. All Certification courses are assessed internally for 100 marks.
- (ii) The Course coordinator handling the course will assess the student through attendance and MCQ test, and declare the student as "pass" on satisfactory completion. A letter grade "P" is awarded to declare pass.
- (iii) The marks scored in these courses will not be taken into consideration for the SGPA / CGPA calculations in the grade sheet.

Evaluation Methods

Assessment	Continuous Assessme	nt Marks (CAM)	Total Marks
Assessment	Attendance	MCQ Test	Total marks
Marks	10	90	100



Department	Artific	ial Intelli	gence and Data Science	Progran	nme: B.	Tech.					
Semester	V			Course	Catego	ry Code:	MC	*End	Semeste	r Exam Ty	pe: -
Course Code	U23A	ADM505		Perio	ods/Wee	ek	Cr	edit	Max	imum Mar	ks
Course Code				L	Т	Р	()	CAM	ESE	TM
Course Name		ENCE OF WLEDGE	INDIAN TRADITIONAL	2	0	0	•	-	100	-	100
	Ţ		Common	to ALL Bra	anches						
Prerequisite	-										
		ompletio	on of the course, the stud	ents will b	e able	to				BT Ma (Highest	
	CO1	Familiar	ize with the philosophy of Indi	an culture						K	2
Course	CO2	Distingu	ish the Indian languages and	literature						K	2
Outcomes	CO3	Describe	e the philosophy of ancient, m	edieval and	modern	India				K	2
	CO4	Illustrate	the information about the fine	arts in Indi	a					K	2
	CO5	Describe	e the contribution of scientists	of different of	eras					K	2
UNIT- I	Intro	oduction ⁻	Γο Culture						Perio	ds:06	
			eritage, general characteristi dia, Modern India	cs of culture	e, import	ance of c	ulture	in hun	nan literati	ıre, Indian	CO1
UNIT- İI	India	an Langu	ages, Culture and Literature						Perio	ds:06	
			- I: the role of Sanskrit, signifi uth India Indian Languages an							hies, other	CO2
UNIT- III			Philosophy						······································	ods:06	····
Religion and Ph India (selected n		•	nt India, Religion and Philoso	phy in Medi	eval Indi	a, Religiou	us Re	form M	ovements	in Modern	СОЗ
UNIT- IV	Fine	Arts in Ir	ndia (Art, Technology and E	ngineering))				Peri	ods:06	
	cient, m		s, Music, divisions of Indian c nd modern), Science and Te								CO4
UNIT-V		cation Sy	stem in India						Perio	ds:06	
			nd modern India, aims of edu edieval India, Scientists of Mo		ects, lar	nguages, S	Scienc	e and	Scientists	of Ancient	CO5

Reference Books

- 1. Kapil Kapoor, "Text and Interpretation: The India Tradition", ISBN: 81246033375, 2005
- "Science in Samskrit", Samskrita Bharti Publisher, ISBN 13: 978-8187276333, 2007
- NCERT, "Position paper on Arts, Music, Dance and Theatre", ISBN 81-7450 494-X, 200
- No Entry, Trodition paper on Aid, Masie, Baries and Thodato, 1987 (1987).
 S. Narain, "Examinations in ancient India", Arya Book Depot, 1993
 M. Hiriyanna, "Essentials of Indian Philosophy", Motilal Banarsidass Publishers, ISBN 13: 978 8120810990, 2014

Web References

- 1. https://nptel.ac.in/courses/109/104/109104102/
- 2. https://nptel.ac.in/courses/101/104/101104065/
- 3. https://nptel.ac.in/courses/109/108/109108158/
- https://nptel.ac.in/courses/109/106/109106059/
 https://nptel.ac.in/noc/courses/noc17/SEM1/noc17-ae01/

COs/POs/PSOs Mapping

COs					Prog	gram O	utcome	s (POs)					ram Spe omes (P	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	1	-	-	-	-	-	-	-	-	3	-	1	1	-	1
2	1	-	-	-	-	-	-	-	-	3	-	1	1	-	1
3	1	-	-	-	-	-	-	-	-	3	-	1	1	-	1
4	1	-	-	-	-	-	-	-	-	3	-	1	1	-	1
5	1	-	-	-	-	-	-	-	-	3	-	1	1	-	1

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

	Continu	ous Assessment N	Marks (CAM)	
Assessment	Attendance	MCQ Test	Presentation / Activity / Assignment	Total Marks
Marks	10	30	60	100

5- NS \-

Department	Artifi	icial Intel	ligence and Data Science	Progran	nme: E	3.Tech.				
Semester	VI			Course	Catego	ory Cod	e: PC Eı	nd Semest	er Exam 7	Гуре: Т Г
Course Code	U23A	ADTC02		Perio	ds/We	ek	Credit	Ма	ximum M	arks
	020 ,			L	Т	Р	С	CAM	ESE	TM
Course Name	NLP	AND CH	АТВОТ	3	0	0	3	25	75	100
	.i		Common to A	AI & DS, (CSE &	BS		·		i
Prerequisite	Machi	ine Learni	ing, Deep Learning and Pro	gramming	j in Pyt	hon				
	On c	ompletio	n of the course, the stude	nts will b	e able	to			(⊢	Mapping lighest ₋evel)
	CO1	Interpret f	fundamental concepts of NLP a	ind apply t	hem for	text pro	cessing			K2
Course	CO2	Apply diff	erent parsing techniques for sy	ntactic and	d semar	tic anal	ysis			K3
Outcomes	CO3	Apply ma	chine translation techniques for	r summariz	zing text	t and qu	estion answe	ring.		K3
	CO4	Understa chatbots	nd the structure and technology	/ behind h	uman-co	omputer	conversation	ns for buildin	g	K2
	CO5	Determin	e various techniques to build a	conversati	onal int	erface	,			K3
UNIT-I	Intro	duction					Periods:	9		
	c ideas i	using Proba	tional semantics, Classical Pars abilistic Context Free Grammars	-		-	-		-	:
UNIT-III	Mach	nine Tran	slation				Periods:	9		i
- Information Extra potential of using UNIT-IV Chatbot – Design nterface, Convers	Chate of a Chate sational	- Introduction dvanced Labot bot atbot - Intro Interface a	chniques, Statistical Machine T on to Named Entity Recognitio anguage Modelling – Application oduction to Conversational Inter and devices - Technology of Co	n and Rela	ation Ex narizatio	traction on, ques	- Natural Lartion answerin Periods: loping a spee	nguage Gen g.) ech based C	eration - th	nal CO 2
of Conversation -	The lan	iguage of C	Conversation.				-			
UNIT-V			al Interface	ontine T	^	`noo-b	Periods:		Cunth:	
mplementing Sp	eech R	ecognition	versational Interface - Implem - Language Model, Acoustic gn – Advanced Chatbots.							
Lecture Perio	ds:45		Tutorial Periods:-	Practic	al Peri	ods:-		TotalPerio	ds:45	i
Text Books		i		4			i			
		_	age Understanding", 3rd Editio							
	anam, "l		Chatbots and Conversational	UI Develo _l	oment:	Build ch	atbots", Publ	ished by Pa	cket Publi	shing Lt

- Srini Janarthanam, "Hands-On Chatbots and Conversational UI Development: Build chatbots", Published by Packet Publishing Ltd., Second Edition, 2020.
- 3. Daniel Jurafsky and James H Martin," Speech and Language Processing: An introduction to Natural Language Processing, Computational Linguistics and Speech Recognition", Prentice Hall, 3rd Edition, 2024.
- 4. Philipp Koehn "Neural Machine Translation" to Neural Machine Translation, Cambridge University Press, 1st edition, 2020.

2-73/-

Reference Books

- Sohom Ghosh, Dwight Gunning, "Natural Language Processing Fundamentals", Packt Publishing Ltd., 1st edition, 2019.
- 2. Jacob Eisenstein," Introduction to Natural Language Processing", MIT Press, 1st Edition, 2019.
- 3. Cathy Pearl, "Designing Voice User Interfaces: Principles of Conversational Experiences", Shroff/O'Reilly, First Edition, 2017.
- 4. Abhishek Singh, Karthik Ramasubramanian, Shrey Shivam, "Building an Enterprise Chatbot: Work with Protected Enterprise Data using Open-Source Frameworks", Apress, 2019.
- 5. Michael McTear," Conversational AI: Dialogue Systems, Conversational Agents, and Chatbots", Publishing Springer 1st Edition 2020.

Web References

- 1. https://onlinecourses.nptel.ac.in/noc23_cs45/preview
- 2. https://towardsdatascience.com/
- 3. https://www.geeksforgeeks.org/natural-language-processing-nlp-tutorial/
- 4. https://www.analyticsvidhya.com/blog/2021/02/basics-of-natural-language-processing/

COs/POs/PSOs Mapping

COs					Prog	ram O	utcom	es (PO	s)					ram Spe omes (P	
	P01	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	2	2	2	3	3	2	-	-	-	-	-	-	2	2	2
2	2	3	3	3	2	1	-	-	-	-	-	-	2	2	1
3	2	3	3	2	1	-	-	-	-	-	-	-	2	2	1
4	2	2	3	2	3	2	-	-	-	-	-	-	2	3	1
5	3	2	2	3	3	1	-	-	-	-	-	-	2	2	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Accomment		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	Attendance	Examination (ESE) Marks	Marks			
Marks	1	5	75	100			

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

2. 1 2/-

Department	Artificial Intelligence and Data Science	Progran	nme: B	.Tech.				
Semester	VI	Course	Catego	ry Code	e: PC End	d Semeste	er Exam T	ype: TE
Course Code	U23ADT611	Perio	ds/Wee	ek	Credit	Ma	ximum Ma	arks
Course Code	UZSADIOII	L	Т	Р	С	CAM	ESE	ΤМ
Course Name	ROBOTIC PROCESS AUTOMATION – UI PATH	3	0	0	3	25	75	100
	Al	l & DS	.4	<u> </u>		<u>i</u>		
Prerequisite	Machine Learning							
	On completion of the course, the stude	nts will b	e able t	o			(H	/lapping ighest evel)
	CO1 Describe RPA, where it can be applied a	and how it's	s implem	ented				K2
Course	CO2 Describe the different types of variables,	Control F	ow and	data ma	nipulation tec	hniques.		K2
Outcomes	CO3 Identify and understand Image, Text and	l Data Tab	les Autoi	mation.				K2
	CO4 Describe how to handle the User Events	and vario	us types	of Exce	otions and str	ategies.		K2
	CO5 Examine the research areas in Artificial I	ntelligence	with res	spect to I	RPA.			K 3
UNIT-I	Introduction To Robotic Process Autom	nation			Periods: 9			
	niques of automation, Robotic process automatio		an RPA	do?, Be			ents of RP	Α,
	The future of automation.			•		, ,		CO1
UNIT-II	RPA Basics				Periods:9			
-	ation - What is RPA - RPA vs Automation - Proc e Automated - Types of Bots - Workloads which o			_	-			
•	PA Development methodologies - Difference from							
	ocess Design Document/Solution Design Documererging ecosystem.	ient - Indu	stries be	st suited	I for RPA - R	sks & Cha	illenges wit	n
UNIT-III	UI Path Introduction and Exploration				Periods:9			
_	alling UiPath Studio community edition - The U	Jser Interf	ace - Ke	evboard		out Updat	tina - Abou	ut
Automation Proje	ects - Introduction to Automation Debugging - Moreome Extension – Variables - Control Flow - D	anaging A	ctivation	Packag	es - Reusing	Automatic	ns Library	- CO3
UNIT-IV	UI Path Advanced Automation				Periods:9			
Image, Text and	Advanced Citrix Automation - Excel Data Tabet Organization. Orchestrator: Tenants – Authent				tomation -De			
UNIT-V	Artificial Intelligence and RPA				Periods:9			L
	lication of RPA for Machine Learning, Agent awa projects on applying RPA for designing and deve				e Processing	- Compute	er Vision, e	tc,
Lecture Perio	ds:45 Tutorial Periods:-	Practic	al Perio	ds:-	Т	otalPerio	ds:45	L
Text Books		<u>I</u>			ii			
1. A. Tripathi	, "Learning Robotic Process Automation: Create UiPath: Create Software robots with the leadingF						s with the	leading

- 2. K. Wibbenmeyer, "The Simple Implementation Guide to Robotic Process Automation (RPA): How to Best Implement RPA in an Organization", iUniverse,2018.
- 3. S. Merianda, "Robotic Process Automation Tools, Process Automation and Their Benefits: Understanding RPA and Intelligent Automation", Createspace.,2018.

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Reference Books

- 1. M. Lacity, L. Willcocks, "Robotic Process and Cognitive Automation: The Next Phase", SteveBrookes Publishing.
- 2. Tom Taulli, "The Robotic Process Automation Handbook: A Guide to Implementing RPASystems", 2020.
- 3. Nandan Mullakara, "Robotic Process Automation Projects: Build real-world RPA solutions usingUiPath and Automation Anywhere", 2020.
- 4. Gerardus Blokdyk, "RPA robotic process automation", Second Edition, Paper Back, 2018.
- 5.S. Muhkerjee, "Essentials of Robotics Process Automation", Khanna Publishing, 2019.

Web References

- 1. https://www.uipath.com/rpa/robotic-process-automation
- 2.https://www.academy.uipath.com
- 3.https://www.edx.org/

COs/POs/PSOs Mapping

COs					Prog	ram O	utcom	es (PO	s)					ram Specific omes (PSOs)	
	P01	PO2	PO3	PO4	PO5	PO6	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	1	2	3	2	1	2	-	-	-	-	-	-	2	3	2
2	1	2	3	2	3	2	-	-	-	-	-	-	3	3	2
3	2	2	2	3	3	1	-	-	-	-	-	-	3	2	3
4	2	2	3	1	3	3	-	-	-	-	-	-	3	3	2
5	2	2	3	2	3	3	-	-	-	-	-	-	3	3	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

_	ssessment		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
^	ssessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
	Marks	1	0	5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

2. 18/ -

Department	Computer Science and Engineering	Progran	nme: B.	Tech				
Semester	III/ V	Course	Catego	ry: PC	End	Semeste	r Exam Tyլ	oe: TE
Course Code	U23CSTC07	Perio	ds/Wee	ek	Credit	Max	kimum Mar	ks
Course Code	023031007	L	Т	Р	С	CAM	ESE	TM
Course Name	WEB DESIGNING	3	-	-	3	25	75	100
	(Common	to CSE and	AI&DS)					
Prerequisite	Basic knowledge in Programming and Databa	ase						
	On completion of the course, the stu	dents will	be abl	e to			BT Ma (Highest	
_	CO1 Understand HTML and CSS						K	2
Course Outcomes	CO2 Implement client-side programming	using JavaS	Script.				K	3
Outcomes	CO3 Understand the concepts of PHP and	d PHP Forn	ns.				K	2
	CO4 Connect PHP scripts with databases	·					K4	4
	CO5 Implement the web hosting processe	es.					K	3
UNIT - I	WEB BASICS, HTML AND CSS				Periods:09)		
Elements.	ax – Location of Styles – Selectors – Box Mod		, ,					
UNIT - II	JAVASCRIPT				Periods:09	}		
_	ւ tion: Syntax – Variables – Operators – Data Tչ	/pes – Fund	ctions –	Objects	 String Meth 	ods – Num	ber Method	S
	ethods – Conditions – Loops – Popup Alert –							
Object Properties –	Object Methods– Object Display.							
UNIT - III	INTRODUCTION TO PHP AND FORM	S			Periods:09)		
Introduction to PHP	: Variables – Data Types – Constants – Echo	/ Print. Op	erators:	Arithmet	tic – Compari	son – Logic	al - String -	-
	witch – Loops – Arrays – Functions – Super glo	•	•			•		CO3
·	/alidation – Form Required – Form Submission	on. Data: D	ate and	Time -	File Upload -	- Cookies -	- Sessions -	_
Include – Exception								
UNIT - IV	PHP WITH DATABASE CONNECTIVIT				Periods:09			
	abase: Essential SQL - Creating a MySQL D							
Database – Access – Sorting the Data.	ing the Database in PHP – Updating Database	s – Insertin	g New D	ata Item	s into a Datab	ase – Delei	ting Record	S CO4
<u> </u>	WED HOSTING				Doriods:00			
UNIT - V	WEB HOSTING	: 0-		:	Periods:09		Danietaria	
	Hosting: Creating the website – Working on the Publishing web sites – Maintaining a website.	ie site – Se	enaing er	nali and	access other	wedsites –	Registering	CO5
	i ubilaning web altea – Maintaining a Website.							

Lecture Periods: 45 Tutorial Periods: - Practical Periods: - Total Periods: 45

Text Books

- 1. Randy Connolly and Ricardo Hoar, "Fundamentals of Web Development", Pearson Education Inc, Third Edition, 2022.
- 2. Steven Holzner, "PHP: The Complete Reference", McGraw Hill Education, 3rd Edition, 2020.
- 3. Jon Dukett, "JavaScript and JQuery: Interactive Front–End Web Development", Paperback, 2018.

Reference Books

- Lyza Danger Gardner, "Java Script on Things: Hacking Hardware for Web Developers", Dreamtech Press,1st edition, 2018.
- 2. Laura Lemay, Rafe Colburn, "Mastering HTML, CSS & Javascript Web", BPB Publications, First edition, 2016.
- Alex Libby, Gaurav Gupta, Asoj Talesra, "Responsive Web Design with HTML5 and CSS3 Essentials", Packt Publishing, 2nd edition, 2016
- 4. Bassett, Lindsay, "Introduction to JavaScript object notation: a to-the-point guide to JSON", O'Reilly Media, 2015.
- 5. Nixon Robin, "Learning PHP, MySQL & JavaScript: With jQuery, CSS & HTML5", O'Reilly Media, 5th edition, 2018.



Web References

- 1. https://developer.mozilla.org/en-US/docs/Learn
- 2. https://www.w3schools.com/sql/default.asp
- 3. https://www.smashingmagazine.com/2021/03/complete-guide-accessible-front-end-components/
- 4. https://alistapart.com/article/mobile-first-css-is-it-time-for-a-rethink/
- 5. https://css-tricks.com/tag/view-transitions/
- 6. https://www.tutorialspoint.com/php/php_introduction.html

* TE – Theory Exam, LE – Lab Exam

COs/POs/PSOs Mapping

	<i>75</i> / 1 O 5	,,	۳.۰۰۰	<u> </u>			T								
COs					Pro	gram	Outcor	nes (P	Os)				Program Specific Outcomes (PSOs)		
	PO1	PO2	PO3	PO12	PSO1	PSO2	PSO3								
1	3	2	1	1	2	-	-	-	-	-	-	-	-	1	2
2	3	2	1	1	2	-	-	-	-	-	-	-	-	1	2
3	2	2	3	-	2	1	-	1	-	-	-	-	-	1	2
4	2	2	3	2	2	2	-	2	-	-	-	-	-	1	2
5	2	2	3	1	-	1	-	2	-	-	-	-	-	1	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High Evaluation Methods

		Con	tinuous Assess	sment Marks (CA	M)	End	
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Semester Examination (ESE) Marks	Total Marks
Marks	5	5	5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

5. X } \ __

	Artificial Intelligence and Data Science	Pro	gran	nme: B.T	ech.			
Semester	VI	Cou	ırse	Category	: PC		End Semes Type: TE	ster Exam
		Per	riods	s / Week	Credit			m Marks
Course Code	U23ADB602	L	Т	Р	С	CAM	ESE	TM
Course Name	BLOCKCHAIN AND CRYPTOGRAPHY	2	0	2	3	50	50	100
	А	\I & [os '		.i	.i	.i	
Prerequisite	Basic Networking Concepts and Good N	Nath	ema	tical Skill	S			· •
Course Outcomes	On completion of the course, the stud	ents	wil	l be able	to			BT Mappin (Highes Level)
	CO1 Describe the use of distributed system	ns in	bloc	kchain tec	hnology and v	vrite smart	contracts	K2
	CO2 Describe the technologies behind cry							K2
	CO3 Acquire knowledge on standard algor authenticity.	ithms	s use	ed to provi	de confidentia	ity, integrity	y, and	K2
	CO4 Apply and Implement the Blockchain currency Wallet.	Conc	epts	and creat	ting basic bloc	ks, Markle	tree, Crypto-	K 3
	CO5 Apply and implement the algorithms D	ES, I	RSA	and Diffie	-Hellman.			K3
UNIT-I	Introduction to Blockchain Technology	v			Periods: 10)		
	Cryptocurrency (Bitcoin) tion – Transactions – Structure - Transactions t	ypes	_ Th		Periods: 10			
	Wallets and its types— Bitcoin payments— Bitcoin programming and the command-line interfa		nves	tment and	I buying and s	elling bitco		
	oin programming and the command-line interfa Cryptography Techniques and Authen	ice –	nves Bitco	tment and	I buying and s	elling bitco		
nstallation – Bitc UNIT-III	oin programming and the command-line interfa	ce – ticat	nves Bitco ion	tment and	buying and sement propose Periods: 10	elling bitco als (BIPs).	ins – Bitcoin	
UNIT-III Symmetric Key E - Public Key Cry	Cryptography Techniques and Authen Algorithms ncryption- Simple DES, Linear and Differential ptography - Factorization problem and RSA, D	ticat	nves Bitco ion analy	tment and oin improve ysis, DES,	Periods: 10 Modes of ope	elling bitco als (BIPs). ration, Trip	ins – Bitcoin	CO3
UNIT-III Symmetric Key E - Public Key Cry Authentication Al	Cryptography Techniques and Authen Algorithms ncryption- Simple DES, Linear and Differential ptography - Factorization problem and RSA, Dgorithms: Message Digest- SHA-1, MD5.	ticat	nves Bitco ion analy	tment and oin improve ysis, DES,	Periods: 10 Modes of ope exchange, Ellip	elling bitco als (BIPs). ration, Trip otic curve c	ins – Bitcoin	CO3
UNIT-III Symmetric Key E - Public Key Cry Authentication Al	Cryptography Techniques and Authen Algorithms ncryption- Simple DES, Linear and Differential ptography - Factorization problem and RSA, Dgorithms: Message Digest- SHA-1, MD5. Laboratory Exercises	ticat	nves Bitco ion analy	tment and oin improve ysis, DES,	Periods: 10 Modes of ope	elling bitco als (BIPs). ration, Trip otic curve c	ins – Bitcoin	CO3
UNIT-III Symmetric Key E Public Key Cry Authentication Al UNIT-IV Creating	Cryptography Techniques and Authen Algorithms ncryption- Simple DES, Linear and Differential ptography - Factorization problem and RSA, Dgorithms: Message Digest- SHA-1, MD5.	ticat	nves Bitco ion analy	tment and oin improve ysis, DES,	Periods: 10 Modes of ope exchange, Ellip	elling bitco als (BIPs). ration, Trip otic curve c	ins – Bitcoin	CO3
UNIT-III Symmetric Key E Public Key Cry Authentication Al UNIT-IV Creating Creation	Cryptography Techniques and Authen Algorithms ncryption- Simple DES, Linear and Differential ptography - Factorization problem and RSA, Dgorithms: Message Digest- SHA-1, MD5. Laboratory Exercises Merkle tree	ticat	nves Bitco ion analy	tment and oin improve ysis, DES,	Periods: 10 Modes of ope exchange, Ellip	elling bitco als (BIPs). ration, Trip otic curve c	ins – Bitcoin	CO3
UNIT-III Symmetric Key E Public Key Cry Authentication Al UNIT-IV Creating Creation Implement	Cryptography Techniques and Authen Algorithms ncryption- Simple DES, Linear and Differential ptography - Factorization problem and RSA, Digorithms: Message Digest- SHA-1, MD5. Laboratory Exercises Merkle tree of Block entation of blockchain in Merkle Trees entation of peer-to-peer network using block ch	ticat crypta	nves Bitco ion analy	tment and oin improve ysis, DES,	Periods: 10 Modes of ope exchange, Ellip	elling bitco als (BIPs). ration, Trip otic curve c	ins – Bitcoin	CO3
UNIT-III Symmetric Key E Public Key Cry Authentication Al UNIT-IV Creating Creation Implement Implement Creating	Cryptography Techniques and Authen Algorithms ncryption- Simple DES, Linear and Differential ptography - Factorization problem and RSA, Digorithms: Message Digest- SHA-1, MD5. Laboratory Exercises Merkle tree of Block entation of blockchain in Merkle Trees entation of peer-to-peer network using block ch	ticat crypta	nves Bitco ion analy	tment and oin improve ysis, DES,	Periods: 10 Modes of ope exchange, Ellip	elling bitco als (BIPs). ration, Trip otic curve c	ins – Bitcoin	CO3
UNIT-III Symmetric Key E Public Key Cry Authentication Al UNIT-IV Creating Creation Implement Implement Creating UNIT-V	Cryptography Techniques and Authen Algorithms ncryption- Simple DES, Linear and Differential ptography - Factorization problem and RSA, Digorithms: Message Digest- SHA-1, MD5. Laboratory Exercises Merkle tree of Block entation of blockchain in Merkle Trees entation of peer-to-peer network using block chain a Crypto-currency Wallet Laboratory Exercises	ticat crypta	ion analy Hellr	tment and bin improv ysis, DES, nan Key E	Periods: 15 Periods: 15 Periods: 15	elling bitco als (BIPs). ration, Trip otic curve c	le DES, AES	CO3
UNIT-III Symmetric Key E Public Key Cry Authentication Al UNIT-IV Creating Creation Impleme Impleme Creating UNIT-V Impleme Substitu Impleme Impleme Substitu Impleme Applythe of the pa	Cryptography Techniques and Authen Algorithms ncryption- Simple DES, Linear and Differential ptography - Factorization problem and RSA, Digorithms: Message Digest- SHA-1, MD5. Laboratory Exercises Merkle tree of Block entation of blockchain in Merkle Trees entation of peer-to-peer network using block chain a Crypto-currency Wallet Laboratory Exercises entation of the following cipher techniques to pation Cipher, c) Hill Cipher. entation of DES algorithm logic. entation of RSA Encryption algorithm a Diffie-Hellman Key Exchange mechanism using arties (Alice) and the JavaScript application as arties (Alice) and the JavaScript application as	ticat crypta crypta ain oerfor	ion analy Helln	tment and bin improven the second sec	Periods: 15 Periods: 15 Periods: 15 Periods: 15	elling bitco als (BIPs). ration, Trip otic curve c	ins – Bitcoin le DES, AES ryptography.	CO4
UNIT-III Symmetric Key E Public Key Cry Authentication Al UNIT-IV Creating Creation Impleme Impleme Creating UNIT-V Impleme Substitu Impleme Impleme Substitu Impleme Applythe of the pa	Cryptography Techniques and Authen Algorithms ncryption- Simple DES, Linear and Differential ptography - Factorization problem and RSA, Digorithms: Message Digest- SHA-1, MD5. Laboratory Exercises Merkle tree of Block entation of blockchain in Merkle Trees entation of peer-to-peer network using block chain a Crypto-currency Wallet Laboratory Exercises entation of the following cipher techniques to particular to provide the company of the properties of the company of the properties of the company of the properties of the pro	ain ing H other	ion ion Helln Tm e	tment and bin improven the second sec	Periods: 15 Periods: 15 Periods: 15 Periods: 15 Periods: 15 According to the periods: 15 Periods: 15 Periods: 15	elling bitco als (BIPs). ration, Trip otic curve c n: a) Caesa ler the end	ins – Bitcoin le DES, AES ryptography.	CO4

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Textbooks

- 1. S. Shukla, M. Dhawan, S. Sharma, S. Venkatesan, Blockchain Technology: Cryptocurrency and Applications, Oxford University Press 2019.
- 2. Van Haren Publishing (Editor), "Introduction to Blockchain Technology: The Many Faces of Blockchain Technology in the 21st Century", Paperback Import, 2019.
- 3. Adrian Mcnulty, "Blockchain: The Complete and Comprehensive Guide to Understanding Blockchain Technologies", Createspace Independent Pub, 2018.
- 4. William Stallings, "Cryptography and Network Security Principles and Practices", Pearson Education; Seventh edition, 2017.

Reference Books

- 1 Arvind Narayanan, Joseph Bonneau, Edward Felten, Andrew Miller, and Steven Gold Feder, Bitcoin and cryptocurrency technologies: a comprehensive introduction. Princeton University Press, 2016.
- 2. William Mougayar, "Business Blockchain Promise, Practice and Application of the Next Internet Technology", John Wiley & Sons 2016.
- 3. Don, Alex Tapscott, "Blockchain Revolution". Portfolio Penguin 2016.
- 4. Wade Trappe and Lawrence C. Washington, Introduction to Cryptography with Coding Theory" Second Edition, Pearson Education, 2007.
- 5. Atul Kahate, Cryptography and Network Security, 2nd Edition, Tata McGraw Hill, 2008.

Web References

- 1.https://www.tutorialspoint.com/cryptography/cryptography_need_for_encryption.htm
- 2. https://www.simplilearn.com/tutorials/blockchain-tutorial
- 3. https://www.javatpoint.com/blockchain-tutorial
- 4. https://www.geeksforgeeks.org/difference-between-rsa-algorithm-and-dsa/
- 5. https://www.includehelp.com/cryptography/digital-signature-algorithm-dsa.aspx

COs/POs/PSOs Mapping

COs				- 4- 4-		gram C	Outcom	es (PO	s)					gram Spe omes (P	
	PO1	PO2	PO3	PO4	PO5	PO6	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	-	-	-	-	-	2	3	2	2	•	•	3	•	-	-
2	-	-	-	-	-	3	3	2	2	-	-	3	-		-
3	-	-	-	-	ı	3	3	2	2	•	1	3	•	•	ı
4	-	-	-	-	•	2	3	2	2		•	3	•	•	-
5	-		-	-	-	2	3	2	2	-	-	3	-	-	-

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

			Con	tinuous Asse	ssment	Marks (CAM) -	- Maximuı	n 50 M	arks			
	C	ontinuc	us Asse	ssment (Theo	ry)	Conti	nuous As	sessm	ent (Pra	ictical)	#En.d	
Assessment	CAT1 CAT2 Model Atter		Attendance	Total	Conduction of Practical	Report	Viva	Total	#End Semester Examination (ESE) Marks (Practical- Internal Evaluation)	Examination	Total Marks	
Marks	5	5	5	5	20*	15	10	5	30*	20	75**	10 0
	*To be	weighte	ed for 10	Marks	10	*To be weight	ed for 10	Marks	10	30	*To be weighted for 50 Marks	

2. 1 / / -

Department	Artifi	cial Intelligence and Data Science								
Semester	VI		Course	Category Code: PC End Semester Exa Type: LE						
Course Code	11234	DP610	Perio	ds / Wee	ek	Credit	Maximum Marks			
Course Code	UZUA	010	L	Т	Р	С	CAM	ESE	ΓM	
Course Name	NLP	AND CHATBOT LABORATORY	0	0	2	1	50	50	100	
	.i	AI &	k DS	.44				. <u>i</u>		
Prerequisite	Machi	ne Learning, Deep Learning and Progra	amming in F	ython						
	On co	ompletion of the course, the students	s will be ab	le to					apping st Level)	
Course	CO1	Apply and develop machine learning mode	els for NLP ir	n Python				K	3	
Outcomes	CO2	Apply NLP techniques to improve informat	tion retrieval.					K	3	
	CO3	Apply and build predictive models.						K	(3	
	CO4	Apply various machine learning algorithms	K3							
	CO5	Create and implement chatbots and OCR	models					K3		

List of Experiments

- 1. Design an application for Sentiment Analysis Using Machine Learning
- 2. Implementation of Resume Screening using Python
- 3. Creation of Named Entity Recognition using spacy
- 4. Implement an information retrieval system using cosine similarity and word embeddings (Word2Vec or GloVe) to match user queries to relevant documents.
- 5. Create a language model using n-grams or neural networks (e.g., LSTM or GPT) to predict the next word in a sequence.
- 6. Compare and evaluate various machine learning algorithms (e.g., Naive Bayes, SVM, Random Forest) for NLP tasks like sentiment analysis, text classification, or NER.
- 7. Create a rule-based or ML-based chatbot using frameworks like Rasa or NLTK, capable of holding basic conversations with users.
- 8. Develop an OCR system using Python's DocTR library to extract text from images or scanned documents.
- 9. Create a rule-based or ML-based chatbot using frameworks like Rasa or NLTK, capable of holding basic conversations with users.
- Implement a speech-to-text (STT) and text-to-speech (TTS) system using Python libraries like Google Speech API or pyttsx3.

Lecture Periods: - Tutorial Periods: Practical Periods: 30 Total Periods: 30

Reference Books

- 1. "Natural Language Processing with Python: Analyzing Text with the Natural Language Toolkit", Steven Bird, Ewan Klein, and Edward Loper ,2nd Edition (2023)
- 2. Deep Learning with Python",2nd Edition (2021), François Chollet.
- 3. "Building Chatbots with Python: Using Natural Language Processing and Machine Learning" by Sumit Raj,1st Edition (2019)
- 4. "Python Machine Learning: Machine Learning and Deep Learning with Python, scikit-learn, and TensorFlow 2".,3rd Edition (2022), Sebastian Raschka and Vahid Mirjalili

Web References

- https://realpython.com/
- 2. https://www.analyticsvidhya.com/blog/2021/06/nlp-application-named-entity-recognition-ner-in-python-with-spacy/
- 3. https://realpython.com/python-nltk-sentiment-analysis/
- 4. https://realpython.com/build-a-chatbot-python-chatterbot/
- 5. https://nanonets.com/blog/ocr-with-tesseract/

COs/POs/PSOs Mapping

COs					Prog	ram O	utcom	es (PO	s)				Program Specific Outcomes (PSOs)			
	PO1	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3	
1	3	2	3	-	-	1	-	-	-	-	-	-	3	2	-	
2	3	2	3	-	-	1	-	-	-	-	-	-	3	2	-	
3	3	2	3	-	-	1	-	-	-	-	-	-	3	2	-	
4	3	2	3	-	-	1	-	-	-	-	-	-	3	2	-	
5	3	2	3	-	-	1	-	ı	1	ı	ı	-	3	2	-	

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Assessment	Con	itinuous <i>A</i>	ssess	ment Marks (C	AM)		
Assessment	Performano cla	e in pract	tical	Model		End Semester Examination	Total Marks
	Conduction of practical	Record work	viva	Practical Examination	Attendance	(ESE) Marks	IVIAI KS
Marks	15	5	5	15	10	50	100

2. 18/ -

Department	Artificial Intelligence and Data Science	Prog	ramme:	B.Tech.				
Semester	VI	Cour	se Cate	gory Cod	de: PC	End Ser	nester Exa	am
						Type: L	.E	
Course Code	U23ADP611	Pe	eriods / \	Week	Maximum Marks			
Course Code	OZGADI OTT	L	Т	Р	С	CAM	ESE	TM
Course Name	ROBOTIC PROCESS AUTOMATION – UI PATH LABORATORY	0	0	2	1	50	50	100

AI & DS

Prerequisite	UI Patl	h Tools	
Course	On co	ompletion of the course, the students will be able to	BT Mapping (Highest Level)
Outcomes	CO1	Implementation of RPA – UI path	K3
	CO2	Develop an application for web scraping, data mitigation and entry process.	К3
	CO3	Apply query processing in email and customer support emails.	K3
	CO4	Develop credit card applications.	K3
	CO5	Apply the automation process in excel and pdf.	K3

List of Exercises

- 1. Extract data from Google Contacts using an API and store it in a structured file format (e.g., CSV, Excel).
- 2. Extract data from an Excel file based on specific conditions and store the results in a new Excel file.
- 3. Segregate emails based on email IDs and organize them into respective folders within Outlook using automation.
- Extract data from various invoice documents, store the data in an Excel file, and automatically send an email to the specified email addresses.
- 5. Develop a system to send automated replies to emails that contain specific text in the subject line.
- 6. Automate the scheduling and processing of daily financial reports using UiPath to streamline reporting for a company.
- 7. Automate the approval process of corporate expense reports using UiPath to enhance efficiency in expense management.
- 8. Automate the process of credit card application processing, including validation and approval workflows, using UiPath.
- Automate the calculation of employee payroll by extracting data from Excel sheets, applying business rules, and generating the final payroll.
- Extract data from PDF invoices and automate the process of storing this information in an organized format (e.g., Excel) using UiPath.

Lecture Periods: Tutorial Periods: Practical Periods: 30 Total Periods: 30 Reference Books

- 1.A. Tripathi, "Learning Robotic Process Automation: Create Software robots and automatebusiness processes with the leading RPA tool UiPath: Create Software robots with the leading RPA tool UiPath", Packt Publishing, 2018.
- 2.K. Wibbenmeyer, "The Simple Implementation Guide to Robotic Process Automation (RPA): How to Best Implement RPA in an Organization", iUniverse,2018.
- 3.S. Merianda, "Robotic Process Automation Tools, Process Automation and Their Benefits: Understanding RPA and Intelligent Automation", Createspace.,2018.
- 4.M. Lacity, L. Willcocks, "Robotic Process and Cognitive Automation: The Next Phase", Steve Brookes Publishing.
- 5.Tom Taulli, "The Robotic Process Automation Handbook: A Guide to Implementing RPA Systems", 2020.

Web References

- 1.https://www.edureka.co/blog/rpa-projects
- 2.https://www.edureka.co/blog/uipath-automation-examples
- 3.https://mindmajix.com/30-rpa-examples

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COs/POs/PSOs Mapping

COs					Prog	ram Oı	utcom	es (PO	s)				Program Specific Outcomes (PSOs)			
	PO1	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3	
1	2	3	2	2	2	-	-	-	-	-	-	-	2	3	3	
2	2	3	3	2	2	-	-	-	-	-	-	-	3	3	3	
3	2	2	3	2	2	-	-	-	-	-		-	2	3	3	
4	3	2	2	2	3	-	-	-	-	-	-	-	3	3	3	
5	2	3	2	2	3	-	-	-	-	-	-	-	3	3	3	

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

	Cor	ntinuous <i>A</i>	Assess	ment Marks (C/	AM)		
Assessment	Performand cla	ce in pract isses	tical	Model		End Semester Examination	Total Marks
	Conduction of practical	Record work	Viva	Practical Examination	Attendance	(ESE) Marks	ivial KS
Marks	15	5	5	15	10	50	100

2. 18/ -

Department	Computer Science and Engineering Programme: B.Tech.										
Semester	III/ V	Course	Catego	ry: PC	End	d Semeste	r Exam Ty	ype: P			
0	U23CSPC06	Perio	ds/Wee	ek	Credit	Max	kimum Ma	ırks			
Course Code	U23C3PC06	L	Т	Р	С	CAM	ESE	TM			
Course Name	e Name WEB DESIGNING LABORATORY 0 0 2 1 50										
	(Common to	CSE and	AI&DS)				<u> </u>			
Prerequisite	Basic knowledge in Programming and Databa	ise									
	On completion of the course, the student	s will be a	ble to				BT Ma (Highes	apping st Level)			
	CO1 Construct and display webpage with H	ITML and C	SS elen	nents			K	(3			
	CO2 Implement JavaScript programming fo	r website cı	eation				K	(3			
Course	CO3 Design PHP Forms K3										
Outcome	CO4 Implement Database connectivity using	g PHP					K	(3			
S	CO5 Web hosting PHP applications.										

List of Exercises

- 1. (a) Design a home page which displays information about your college department using headings, HTML entities and paragraphs.
 - (b) Create a webpage for any clinic using marquee and HTML formatting tags.
- 2. Design a timetable and display it in tabular format.
- Design an admission form for any course in your college with text, password fields, drop-down list, checkboxes, radio buttons, submit and reset button etc.
- 4. Design a web page of your home town with an attractive background color, text color, an image, font face by using Inline CSS formatting.
- 5. (a) Design a web page by using different CSS border styles.
 - (b) Demonstrate the use of CSS Box Model.
- 6. Write a JavaScript program to remove a character at the specified position of a given string and return the new string.
- Develop and demonstrate a HTML file that includes JavaScript script for taking a number n as input using prompt and display first n Fibonacci numbers in a paragraph.
- 8. Design HTML form for keeping student record, apply JavaScript validation in it for restriction of mandatory fields, numeric field, email-address field, specific value in a field etc.
- 9. Write a program in PHP for processing a simple form (use controls like checkbox, radio buttons and options).
- 10. Write a program in PHP for a simple POST and GET functions
- 11. Design a login form using cookies, bootstrap, PHP, Database.
- 12. Design a student form with add, update, delete, display all and search option using student database.

12. 200.g. a o.a	40	min ada, apadio, aoio	,,	nay an and obaron option doin	g olddolli ddidbdool	
Lecture Periods:	-	Tutorial Periods:	-	Practical Periods:30	Total Periods:30	
Poforonco Books						

- 1. Nixon Robin, "Learning PHP, MySQL & JavaScript: With jQuery, CSS & HTML5", O'Reilly Media, 5th edition, 2018.
- 2. Lyza Danger Gardner, "Java Script on Things: Hacking Hardware for Web Developers", Dreamtech Press, 1st Edition, 2018.
- 3. Steven Suehring, Janet Valade, "PHP, MySQL, JavaScript & HTML5 All-in-One", John Wiley and Sons Inc, 2017.
- 4. Keith Wald, Jason Lengstorf, "Pro PHP and jQuery", Paperback, 2016.
- 5. Laura Lemay, Rafe Colburn, "Mastering HTML, CSS and Javascript Web", BPB Publications, First edition, 2016.

Web References

- 1. https://www.w3schools.com/php/DEFAULT.asp
- 2. https://www.tutorialspoint.com/php/index.html
- 3. https://www.phptpoint.com/php-tutorial/
- 4. https://www.javatpoint.com/php-tutorial
- 5. https://www.w3schools.com/html/default.asp

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COs/POs/PSOs Mapping

Cos					Pro	gram	Outcor	nes (P	Os)				Program Specific Outcomes (PSOs)			
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3	
1	3	3	3	3	3	3	3	3	-	2	3	-	3	3	3	
2	3	3	3	3	-	3	-	3	-	2	-	2	2	2	-	
3	2	2	2	2	2	2	3	3	-	3	3	-	2	2	2	
4	2	2	2	2	2	2	-	3	-	3	-	3	3	3	-	
5	3	3	3	3	3	3	3	3	-	3	3	3	3	3	3	

Correlation Level: 1-Low, 2-Medium, 3-High

Evaluation Method

	Co	ontinuous	Assessm	nent Marks (CAI	M)		
Assessment		nce in prac lasses	tical	Model		End Semester Examination	Total Marks
	Conduction of practical	Conduction Record viva			Attendance	(ESE) Marks	ind No
Marks	15	5	5	15	10	50	100

5. X } \ __

Department	Artificial Intelligence and Data Science	Progran	nme: B	.Tech.								
Semester	VI	Course Category Code: PE End Semester Exam Type: TE										
Course Code	U23ADE610	Perio	ds/We	ek	Credit	Max	ximum	n Marks E TM				
Course Code	UZSADEUTU	L	Т	Р	С	CAM	ESE TI					
Course Name	SPEECH PROCESSING AND ANALYTICS 3 0 0 3 25											
	Al	& DS	4	.ii.			L					
Prerequisite	Natural Language Processing											
	On completion of the course, the students will be able to											
	CO1 Identify the different linguistic component	s of natura	al langu	age				K2				
Course Outcomes	CO2 Interpret a morphological analyzer for a given natural language and apply it for analyzing speech effectively											
	CO3 Decide on the appropriate parsing techni	ques nece	essary fo	or a given	language and	d application	on	K2				
	CO4 Apply new tag set and a tagger for a give				K3							
	CO5 Interpret various techniques in speech re	ly them fo	for text to speech conversion			K2						
UNIT-I	Speech Processing				Periods: 9		<u>i</u>					
Log Spectral Dis	Speech Analysis The Extraction and Pattern Comparison Techniques: Stance, Cepstral Distances, Weighted Cepstral Distances, Verguency Scale, LPC, PLP and MFCC Coefficients	tances an	d Filterir	ng, Likelih	nood Distortion	ns, Spectra	al Disto	rtion				
Multiple Time – A				·								
UNIT-III	Speech Modeling				Periods:9							
-	y Continuous Speech Recognition: Architecture on nguage models – n-grams, context dependent sub-	-		-	•	-	n syste	em –				
UNIT-IV	Speech Recognition				Periods:9			<u>i</u>				
	nomorphic Systems for Convolution: Properties of um of Speech, Pitch Detection, Formant Estimation					al Conside	erations	The CO4				
UNIT-V	Speech Synthesis				Periods:9			i				
	Synthesis: Concatenative and waveform synthesis, Applications and present status.	methods,	sub-wo	ord units f	or TTS, intelli	gibility and	l natura	co:				
Lecture Perio	ods:45 Tutorial Periods:-	Practica	al Perio	ods:-	To	otalPerio	ds:45	L				
Lecture Ferro												
Text Books					<u></u>							

- Reference Books
- 1. Saxena, V. N, "Speech Signal Processing: Using Matlab", Khanna Publishing, 2017.
- 2. Vuppala, R. A. O, "Speech Processing in Mobile Environments", Springer, 2018.
- 3. Udayashankara. V, "Modern Digital Signal Processing", PHI Learning, 2017.



Web References

- 1. https://onlinecourses.nptel.ac.in/noc24_ee118/previewhttps://www.electrical4u.com/
- 2. https://www.coursera.org/courses?query=speech%20recognition
- $3. \ https://www.shiksha.com/online-courses/speech-recognition-certification\\$
- 4. https://web.ece.ucsb.edu/Faculty/Rabiner/ece259/speech%20course.h

COs/POs/PSOs Mapping

COs		Program Outcomes (POs)											Program Specific Outcomes (PSOs)		
	P01	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	3	2	2	3	-	-	-	-	-	-	-	3	2	1
2	3	3	3	2	3	-	-	-	-	-	-	-	3	3	2
3	3	3	3	2	3	-	-	-	-	-	-	-	3	2	2
4	3	2	2	2	3	-	-	-	-	-	-	-	3	3	2
5	3	3	3	3	3	-	-	-	-	-	-	-	3	2	3

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Assessment		Continuou	End Semester	Total			
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	10		5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

5- X S / __

Department	Inforr	nation Technology	Prograr	nme: B.	Tech.				
Semester	VIII/V		Course	Catego	ry Code	: PE *Er	nd Semes	ter Exam T	ype: TE
			Peri	ods / We	eek	Credit	Ma	ximum Ma	rķs
Course Code	U2311	EC05	L	Т	Р	С	CAM	ESE	TM
Course Name	Augn	nented Reality and Virtual Reality	3	0	0	3	25	75	100
	(Commo	on to IT, AIDS and ECE)							
Prerequisite	Com	outer Graphics							
	On c	ompletion of the course, the stud	ents will b	e able t	to			BT Ma (Highes	
	CO1	Understand the fundamentals of Virtua	ıl reality					K	2
	CO2	Explain the concepts of motion and tra	cking in VR	systems				K	2
Course Outcome	CO3	Describe the importance of interaction	and audio i	n VR sys	tems			K	2
Outcome	CO4	Understand and work on Augmented F	Reality envir	onment				K	2
	CO5	Explore the application area of augment	nted and vir	tual realit	ty			K	3
Unit- I	Introd	uction				Periods: 0	9	<u>-</u>	
mproving Later	ncy and F	ual Rendering - Ray Tracing and Sha Frame Rates - Motion in Real and Virto Tracking Attached Bodies.							
Unit- III	***************************************	action & Audio				Periods: 0	9		
		ams and Remapping – Locomotion – Nn Hearing - Auditory Perception – Audit							
Unit- IV	Fund	amentals of AR				Periods: 0	9		
		ed Reality – Origin of AR – Definitic epts – AR Content – Interaction – Mobil		elationsh	nip Betw	een Augmer	nted Realit	y and Othe	cO4
Unit- V	Appl	ications of Augmented and Virtua	l Reality			Periods: 0	9		
Applications - Education.	Gaming	and Entertainment - Science and En	gineering -	Health	and Med	dicine - Aero	space and	d Defence	_ CO5
Lecture Perio	ds: 45	Tutorial Periods:	Practic	al Perio	ods: -	Т	otal Perio	ods: 45	
Γext Books									
		Steven M. LaValle, Cambridge University Jnderstanding Augmented Reality, Cond			ns, Morga	an Kaufmann	,		
	al Augma	ented Reality: A Guide to the Technolo	aioo Applia	otiono o	مريكا لمرم	on Footore fo	or AD and	VD (Hookiii	+v) C+ov

3. Practical Augmented Reality: A Guide to the Technologies, Applications, and Human Factors for AR and VR (Usability), Steve Aukstakalnis, Addison-Wesley Professional; 1 edition, 2016

Reference Books

- 1. Burdea, G. C. and P. Coffet. Virtual Reality Technology, Second Edition. Wiley-IEEE Press, 2003/2006
- 2. Alan Craig, William Sherman, Understanding Virtual Reality, Second Edition, Morgan Kaufmann, 2018.
- 3. Augmented Reality: Principles and Practice (Usability) by Dieter Schmalstieg & Tobias Hollerer, Pearson Education (US), Addison-Wesley Educational Publishers Inc, New Jersey, United States, 2016. ISBN: 9780321883575
- Augmented Reality: Principles & Practice by Schmalstieg / Hollerer, Pearson Education India; First edition (12 October 2016), ISBN-10: 9332578494

Web References

- 1. https://nptel.ac.in/courses/121106013
- 2. https://onlinecourses.swayam2.ac.in/nou24_ge37/preview
- 3. http://cambum.net/course-2.htm
- 4. https://www.youtube.com/watch?v=MGuSTAqlZ9Q

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COs/POs/PSOs Mapping

COs					Prog	ram O	utcom	es (PO	s)				Program Specific Outcomes (PSOs)		
	PO1	PO2	PO3	PO4	PO5	PO6	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	3	2	1	2	-	-	-	-	-	-	-	3	2	1
2	2	3	2	2	3	-	-	-	-	-	-	-	3	3	2
3	3	3	3	1	2	-	-	-	-	-	-	-	3	2	1
4	1	2	1	3	3	-	-	-	-	-	-	-	2	2	1
5	2	1	2	3	2	-	-	-	-	-	-	-	1	3	3

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continuou	s Assessmei	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	5	5	5	75	100

^{*}Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

Department	Artifi	icial Intel	ligence and Data Science	Program	nme: B	.Tech.				
Semester	VI			Course	Catego	ry Code	e: PE En	d Semeste	er Exam 7	Гуре: ТЕ
Course Code	11234	ADE611		Perio	ds/Wee	ek	Credit	Ma	ximum M	arks
Course Code	OZOF	(DEUI)		L	Т	Р	С	CAM	ESE	TM
Course Name	ADV	ANCED .	JAVA PROGRAMMING	3	0	0	3	25	75	100
	.1		A	I & DS			<u> </u>			
Prerequisite	Funda	amentals	of Java Programming, Web	Technolo	gy					
	On c	ompletio	n of the course, the stude	nts will b	e able t	to			(⊢	Mapping lighest ₋evel)
	CO1	Interpret	and apply the concepts of Java	utility pacl	kages to	create (data structure	S		K2
Course	CO2	Apply pro	oficiency in data structures using	g Java Col	lection F	ramewo	ork			K3
Outcomes	CO3	Apply Spusing YA	ring framework in Java for build ML	ling enterp	rise Java	a Applica	ations and au	to configure	e it	K3
	CO4	Interpret	various Spring Databases to ha	andle datab	oases at	the bac	kend			K2
	CO5	Interpret	how microservices are helpful i	n commun	ication u	sing AP	I			K2
UNIT-I	Java	Utility P	ackage				Periods: 9)	i	
Introduction to ja Interface – Key c			– Array list – List Interface – H asses	lashMap In	nterface -	– Set Int	terface – Que	ue Interfac	e – Deque	ue CO1
UNIT-II			n Framework				Periods: 9	-		
•	•	~ .	ection – linked list – queue – st	• .			m API and Fu	ınctional Pr	rogrammin	g: CO2
	····γ·········		al Interfaces and Lambda Expre	ession - Op	tional Cl	ass				
UNIT-III	à	ng Frame					Periods: 9			
configuration to sp	•	•	endency injection – inversion o	f control –	bean fac	otory – a	pplication con	itext – cond	cepts of au	CO3
UNIT-IV		ng Datab					Periods: 9			
data JDBC – Spri	ing Boot	t Integration	ງ databases – Spring data acce n – Spring Data REST – Sprinç		g data ad	ccess- S			oDB – spr	ing CO4
UNIT-V			s in Java				Periods: 9			
			y – spring cloud circuit breake sourcing – Rest API – HTTP m							
Lecture Perio	ds:45		Tutorial Periods:-	Practica	al Perio	ds:-	T	TotalPerio	ds:45	
Text Books				.4			<u>i</u>			

- 1. Uttam K. Roy, "Advanced Java Programming", Oxford University Press, 2015
- 2. Claudio Eduardo de Oliveira, Greg L. Turnquist, Alex Antonov, "Developing Java Applications with Spring and Spring Boot", Packt Publishing, 7th edition ,2018.
- 3. John Carnell and Illary Huaylupo Sanchez , "Spring Microservices in Action" , Manning Publications Co ,2nd edition 2021.
- 4. Craig Walls, "Spring in Action", Manning, 5th edition, 2018
- 5. B. Prasanalakshmi , "Advanced Java Programming", CBS Publishers & Distributors, 2015

Reference Books

- 1. Cay. S Horstmann and Gary Cornell, "Core Java: Volume II Advanced Features", Pearson, 12th Edition, 2023.
- 2. Rod Johnson, Juergen Hoeller, Alef Arendsen, Thomas R, "Professional Java Development With The Spring Framework", Wiley India Pvt. Limited, 2009
- 3. Dr.Rajendra Kawale, "Advanced Java", Devraj Publications, Mumbai, 2018
- 4. Holzner, Steven et.al, "Java 2 Programming Black Book", DreamTech Press, New Delhi, 2009
- 5. Herbert Schildt, "Complete Reference Java", Mcgraw Hill Education, New Delhi, 7th Edition, 2021

Academic Curriculum and Syllabi R-2023

Web References

- 1. https://nptel.ac.in/courses/106105084/30
- 2. https://www.javatpoint.com/java
- 3. https://www.tutorialspoint.com/java
- 4. https://javabrains.io/
- 5. https://spring.io/projects/spring-cloud

COs/POs/PSOs Mapping

COs					Prog	ram Oı	utcom	es (PO	s)					ram Spe omes (P	
	PO1	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	2	1	-	-	-	-	-	-	-	1	1	1	2	1	-
2	3	2	1	1	-	-	-	-	-	1	1	1	2	1	-
3	3	2	1	1	-	-	-	-	-	1	1	1	2	1	-
4	3	2	1	1	-	-	1	-	1	1	1	1	2	1	1
5	3	2	1	1	-	-	-	-	-	1	1	1	2	1	-

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Assessment		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	5	75	100		

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

5-18/-

Department	Artifi	cial Intelligence and Data Science	Progran	nme: I	3.Tech.				
Semester	VI		Course	Catego	ory Cod	e: PE E	nd Semester	Exam T	ype: TE
Course Code	1100 A	DEC42	Perio	ods/We	···· ·	Credit		mum Ma	arks
Course Code	UZ3A	.DE612	L	Т	Р	С	CAM	ESE	TM
Course Name	PRE	DICTIVE DATA ANALYTICS	3	0	0	3	25	75	100
Droroguioito	Mathai	AI matics and Statistics, Python or R, Pro	& DS	lvina c	kill Dat	n vicualizat	ion		
Prerequisite		ompletion of the course, the studer				a visualizai	.iOTT	(H	Mapping ghest evel)
	CO1	Interpret basic principles of predictive an	alytics and	d apply	it for buil	ding models	and application		K2
Course Outcomes	CO2	Differentiate various statistical methods u	used for p	rediction	n and eva	aluate mode	l performance		K2
Outcomes	CO3	Apply classification methods for predictive	e modelin	ng					K3
	CO4	Apply Specialized Techniques to forecas	st future tre	ends an	d feature	engineerin	g		K3
	CO5	Apply Model Interpretability and Integrati	on into De	ecision-l	Making				K3
UNIT-I	Intro	duction to Predictive Analytics				Periods	: 9	<u>i</u>	
	standing	forming data into predictive insights - App g the Predictive Analytics workflow: Data e s							
UNIT-II	1	stical Foundations for Prediction				Periods			
Distributions: Und	derstand	concepts: Descriptive statistics, hypother ling data variability for prediction - Linea nance: Metrics like R-squared and Mean S	ar Regres	sion: Bu					
UNIT-III	Class	sification for Predictive Modeling				Periods			
(SVM) for comple	x relatio	dicting binary outcomes (yes/no) - Class nships - Model Selection and Regularizat abining multiple models for improved accu	ion Techn						CO3
UNIT-IV	Adva	nced Predictive Techniques				Periods	: 9		
oetween variables	s - Clust	precasting future trends based on histori pering for Segmentation: Grouping data po ransforming data for better model perform	oints base						
UNIT-V	Deplo	oyment and Impact of Predictive Mo	odels			Periods	: 9		
oerformance over	time - I	nderstanding how models make predictive Models into Decisionations: Exploring trends like Deep Learning	n Making	Proces	ses: Usi				
Lecture Perio	ds: 45	Tutorial Periods: -	Practic	al Peri	ods: -		Total Period	ls: 45	
Text Books									
2. Friedman, J., Edition, 2009	Hastie,	George Athanasopoulos, "Forecasting: F T., & Tibshirani, R." The Elements of Sta chanasopoulos, G. Forecasting: Principles	tistical Lea	arning: l	Data Min	ing, Inferen	ce, and Predict		
Reference Boo									
		non for Data Analysis", Publisher: O'Reilly and Vahid Rostamzadeh, "Machine Learni			Publishe	r: Packt Pub	olishing, Edition	: 2nd, 20	19

- 2. Sebastian Raschka and Vahid Rostamzadeh, "Machine Learning with Python", Publisher: Packt Publishing, Edition: 2nd, 2019
- 3. Bishop, C. M., "Pattern Recognition and Machine Learning", Springer, 2006.
- 4. James, G., Witten, D., Hastie, T., & Tibshirani, R, "An Introduction to Statistical Learning with Applications in R". Springer, 2013.

2-73/-

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Web References

- 1. https://machinelearningmastery.com/tutorial-first-neural-network-python-keras/
- https://www.sas.com/en_in/insights/analytics/predictive-analytics.html
- 3. https://www.kaggle.com/datasets
- 4. https://archive.ics.uci.edu/ml/index.php
- 5. https://data.gov/

COs/POs/PSOs Mapping

COs					Prog	ram O	utcom	es (PO	s)					ram Spe omes (P	
	PO1	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	2	2	2	3	-	-	-	-	-	-	-	3	3	-
2	3	3	3	3	3	-	-	-	-	-	-	-	3	3	-
3	3	3	2	2	3	-	-	-	-	-	-	-	3	3	-
4	3	3	3	3	3	-	-	-	-	-	-	-	3	3	-
5	3	2	2	2	3	-	-	-	-	-	-	-	2	3	-

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Assessment		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	75	100			

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

5- X S/ --

	Artificial Intelligence and Data Science	Program	nme: B	.Tech.				
Semester	VI	Course (Catego	ry Code	: PE End	d Semeste	er Exam T	Гуре: Т І
		Perio	ds/We	ek	Credit	Max	ximum M	arks
Course Code	U23ADE613	L	Т	Р	С	CAM	ESE	TM
Course Name	ADVANCED NATURAL LANGUAGE PROCESSING	3	0	0	3	25	75	100
	A	l & DS						<u>.</u>
Prerequisite	Python and Natural Language Processing							
	On completion of the course, the stude	nts will be	e able	to			(H	Mappin lighest ₋evel)
000000	CO1 Describe the fundamentals of natural la	nguage prod	cessing	and app	ly it for langua	age modelir	ng	K2
Course Outcomes	CO2 Interpret various word level analysis for	syntactic pa	arsing					K2
Catoomico	CO3 Interpret various techniques for handling	g word sens	es and	semantio	CS			K2
	CO4 Apply different and advanced technique	s for handlir	ng disco	ourses ar	nd Lexical Res	sources.		K3
	CO5 Interpret the modern NLP concepts and	apply in it in	n buildir	ng real tii	me application	าร.	K	(2,K3
UNIT-I	Introduction				Periods: 9		<u></u>	
UNIT-II Unsmoothed N-g	Word Level Analysis grams, Evaluating N-grams, Smoothing, Interpolat	ion and Dag	skoff _ V	WI OI-	Periods:9			i
Free Grammars	c and Transformation-based tagging, Hidden Ma , Grammar rules for English, Dynamic Progra	rkov and Ma mming, Pa	aximum	Entropy	models. synta	actic analys	sis-Contex	ct-
Free Grammars	c and Transformation-based tagging, Hidden Ma	rkov and Ma mming, Pa	aximum	Entropy	models. synta	actic analys	sis-Contex	ct-
Free Grammars Lexicalized CFG: UNIT-III Requirements for Word Senses, Supervised, Diction	c and Transformation-based tagging, Hidden Ma, Grammar rules for English, Dynamic Prograss - Feature structures, Unification of feature structures and Pragmatics representation, First-Order Logic, Description Logic Relations between Senses, Thematic Roles, selonary & Thesaurus, Bootstrapping methods – Wo	rkov and Mamming, Pa tures. ogics – Syntectional res ord Similarit	aximum rsing – tax-Driv strictions	Entropy Shallov yen Sema	models. synta v 85 parsing Periods:9 antic analysis d Sense Disar us and Distrib	actic analys —PCFG, F , Semantic mbiguation,	sis-Contex Probabilist attachmen , WSD usi	ic nts
Free Grammars Lexicalized CFG: UNIT-III Requirements for Word Senses, Supervised, Diction	c and Transformation-based tagging, Hidden Ma, Grammar rules for English, Dynamic Progras - Feature structures, Unification of feature structures and Pragmatics representation, First-Order Logic, Description Logic Relations between Senses, Thematic Roles, selonary & Thesaurus, Bootstrapping methods – Wo	rkov and Mamming, Pa mming, Pa tures. ogics – Syntectional res ord Similarity	aximum rsing – tax-Driv strictions y using	Entropy Shallov yen Sema s – Word Thesaur	models. synta v 85 parsing Periods:9 antic analysis, d Sense Disar us and Distrib Periods:9	actic analys —PCFG, F , Semantic mbiguation, outional met	sis-Contex Probabilist attachmen , WSD usi thods.	nts ing CO3
Free Grammars Lexicalized CFG: UNIT-III Requirements for Word Senses, Supervised, Diction UNIT-IV Discourse segme Coreference Resistrown Corpus, B	c and Transformation-based tagging, Hidden Ma, Grammar rules for English, Dynamic Prograss - Feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Programmatics of Feature Structures, Programmatic Programmatics of Programmatics of Structures, Unification of Structures, Unification of Structures, Unification of Structures, Unification of Structures, Unification of Structures, Unification of Structures, Unification of feature structures, Unification of features, Unification of	rkov and Mamming, Pastures. ogics – Syntectional resord Similarity ees Anaphora I	aximum rsing – tax-Driv strictions y using Resolut	Entropy Shallov yen Semas – Word Thesaur	Periods:9 antic analysis d Sense Disar us and Distrib Periods:9 g Hobbs and gger, WordNe	actic analys —PCFG, F , Semantic mbiguation, putional met	attachmer , WSD usi thods.	nts CO:
Free Grammars Lexicalized CFG: UNIT-III Requirements for Word Senses, Supervised, Diction UNIT-IV Discourse segme Coreference Researown Corpus, B UNIT-V	c and Transformation-based tagging, Hidden Ma, Grammar rules for English, Dynamic Prograss - Feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification, Description Legal Research Programmers, Semantics and Pragmatics Relations between Senses, Thematic Roles, selection of the Senses, Thematic Roles, Selection of the Senses, Thematic Roles, Selection of the Senses, Thematic Roles, Selection of the Senses of the	rkov and Mamming, Pa tures. ogics – Syntectional resord Similarity es Anaphora I r, Penn Tree	aximum rsing – tax-Driv strictions y using Resolut ebank, I	Entropy Shallow ven Sema s – Word Thesaur ion using Brill's Tag	Periods:9 antic analysis d Sense Disar us and Distrib Periods:9 g Hobbs and gger, WordNe	actic analys —PCFG, F , Semantic mbiguation, outional mer Centering t, PropBank	attachmee , WSD usi thods. Algorithme	nts CO:
Free Grammars Lexicalized CFG: UNIT-III Requirements for Word Senses, Supervised, Diction UNIT-IV Discourse segme Coreference Research Corpus, B UNIT-V Language encodionit Neural Netvo	c and Transformation-based tagging, Hidden Ma, Grammar rules for English, Dynamic Prograss - Feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Programmatics of Feature Structures, Programmatic Programmatics of Programmatics of Structures, Unification of Structures, Unification of Structures, Unification of Structures, Unification of Structures, Unification of Structures, Unification of Structures, Unification of feature structures, Unification of features, Unification of	rkov and Mamming, Patures. ogics – Symectional resord Similarity es Anaphora If renormal resord of the control of the contr	tax-Driver tax-Driver tax-Driver trictions by using Resolute bank, I be Neural extraction - Missing - Miss	ren Semas – Word Thesaur ion using Brill's Tag	Periods:9 g Hobbs and gger, WordNe Periods:9 g Hobbs and gger, WordNe Periods:9 ks for NLP - I media Information 18 particular	, Semantic mbiguation, putional meronal meta-learni ation Extra	attachmen , WSD usi thods. Algorithm k, FrameN	nts co:
Free Grammars Lexicalized CFG: UNIT-III Requirements for Word Senses, Supervised, Diction UNIT-IV Discourse segments Brown Corpus, Brown Corpu	c and Transformation-based tagging, Hidden Ma, Grammar rules for English, Dynamic Progras s - Feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Boetstrapping Description Legal Relations between Senses, Thematic Roles, set onary & Thesaurus, Bootstrapping methods – Wolf Discourse Analysis and Lexical Resources entation, Coherence – Reference Phenomena, Olution – Resources: Porter Stemmer, Lemmatize ritish National Corpus (BNC). Modern NLP ing and decoding - Multilingual multimedia Encoworks For Information Extraction - Multilingual Intion Extraction - Schema Induction and Knowled e Generation - Knowledge Controlled Language (1997).	rkov and Mamming, Patures. ogics – Symectional resord Similarity es Anaphora If renormal resord of the control of the contr	tax-Drivertictions y using Resolutebank, I	ren Semas – Word Thesaur ion using Brill's Tag	Periods:9 antic analysis, d Sense Disar us and Distrib Periods:9 g Hobbs and gger, WordNe Periods:9 ks for NLP - I media Information detection	, Semantic mbiguation, putional meronal meta-learni ation Extra	attachmer , WSD usi thods. Algorithm k, FrameN ng for NLI ction - Op Answerin	nts co
Free Grammars Lexicalized CFG: UNIT-III Requirements for Word Senses, Supervised, Diction UNIT-IV Discourse segme Coreference Research Corpus, B UNIT-V Language encodice oint Neural Netwood Neural Netwood Neural Language Lecture Periof Text Books	c and Transformation-based tagging, Hidden Ma, Grammar rules for English, Dynamic Progras s - Feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Unification of feature structures, Boetstrapping Description Legal Relations between Senses, Thematic Roles, set onary & Thesaurus, Bootstrapping methods – Wolf Discourse Analysis and Lexical Resources entation, Coherence – Reference Phenomena, Olution – Resources: Porter Stemmer, Lemmatize ritish National Corpus (BNC). Modern NLP ing and decoding - Multilingual multimedia Encoworks For Information Extraction - Multilingual Intion Extraction - Schema Induction and Knowled e Generation - Knowledge Controlled Language (1997).	rkov and Mamming, Patures. ogics – Syntectional resord Similarity es Anaphora If resord Tree ding - Deep formation Ege Acquistic Generation. Practica	tax-Driv strictions y using Resolut ebank, I	ren Semas – Words – Wo	Periods:9 antic analysis, Sense Disarus and Distrib Periods:9 g Hobbs and gger, WordNer Periods:9 ks for NLP - I media Information detection	Actic analys —PCFG, F , Semantic mbiguation, outional mer Centering t, PropBank Meta-learni ation Extra - Question	attachmen attachmen , WSD usi thods. Algorithm k, FrameN or NLI ction - Op Answerin	nts co

- 3. Irum Hafeez Sodhar and Abdul Hafeez Buler, "Natural Language Processing:Applications,techniques and Challenges", Akinik Publications, 2020.

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Reference Books

- 1. Roussanka Loukanova, "Natural Language Processing in Artificial Intelligence", Springer Nature, 2020.
- 2. Akshay kulkarani and Adarsha Shivananda, "Natural Language processing Recipes", Apress, 2019.
- 3. Brian McMahan and Delip Rao,"Natural Language Processing with Pytorch:Build Intelligent Language Applications using Deep Learning", O'Reilly Media, 2019.
- 4. Sowmya Vajjala,Bodhisattwa Majumder,Anuj Gupta and Harshit Surana, "Practical Natural Language Processing",First Edition,O'Reilly Media, 2020
- 5. Lewis Tunstall, Leondro von Werra, Thomas Wolf, "Natural Language Processing with Transformers: Building Language Applications with Hugging Face, Revised Colour Edition", First Edition, Shroff/O.Reilly Media, 2022.

Web References

- 1. https://aws.amazon.com/whatis/nlp/#:~:text=Natural%20language%20processing%20(NLP)%20is,manipulate%2C%20and%20comprehend%20human%20language.https://www.electrical4u.com/
- 2. https://www.tutorialspoint.com/natural_language_processing/natural_language_processing_syntactic_analysis.htm
- 3. https://linguistics.uga.edu/research/content/pragmatics-and-discourse-analysis#:~:text=Pragmatics%20and%20Discourse%20Analysis%20involve,relation%20to%20its%20social%20context.
- 4. https://nlpoverview.com/

COs/POs/PSOs Mapping

COs					Prog	ram Oı	utcom	es (PO	s)					ram Spe omes (P	
	P01	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	2	-	2	2	-	-	-	-	-	-	1	-	2	2
2	3	2	-	2	2	-	-	-	-	-	-	1	-	2	2
3	3	2		2	2	-	•	-	-	-		1	-	2	2
4	3	2	ı	2	2	-	ı	-	1	-	=	1	-	2	2
5	3	2	-	2	2	-	-	-	-	-	-	1	-	2	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Accessment		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	5	75	100		

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

5- X } \ __

Department	Artific	ial Intelligence and Data Science	ligence and Data Science Programme: B. Tech.										
Semester	VI		Cour	se Cate	gory Co	ode: PA	*End Se	emester	Exam Type: -				
Course	LIOOAF	NW602	Pe	riods / V	Veek	Credit		Maximu	ım Marks				
Code	UZSAL	DW602	L	Т	Р	С	CAM	ESE	TM				
Course Name	MINI F	PROJECT	0	0	2	1	100	-	100				
		Α	I & DS										
Prerequisite	Artifi	cial Intelligence, Machine Learning, D	Deep Lo	earning,	Progra	mming in	C, Pytho	n, Java					
	On c	ompletion of the course, the stude	nts wi	ll be ab	le to				BT Mapping (Highest Level)				
Course	CO1	Identify the problem statement for the r	nini proj	ect work	through	the literat	ure survey	,	K2				
Outcomes	CO2	Choose the proper components as per	the req	uirement	s of the	design/ sy	stem.		K2				
	CO3	Apply the acquainted skills to develop f	inal mo	del/syste	m				K3				

There shall be a Mini Project, which the student shall pursue as a team consists of maximum 4 students during the third year, fifth semester. The aim of the mini project is that the student has to understand the real time hardware / software applications. The student should gain a thorough knowledge in the problem he/she has selected and in the hardware / software he/she using in the Project. The Mini-project is an application that should be formally initiated and should be developed and also to be implemented by the respective team.

The Mini Project shall be submitted in a report form along with the hardware model / software developed, duly approved by the department internal evaluation committee. It shall be evaluated for 100 marks as Continuous Assessment. The department internal evaluation committee shall consist of faculty coordinator, supervisor of the project and a senior faculty member of the department. There shall be two reviews that will be considered for assessing a Mini Project work with weightage as indicated evaluation Methods.

Lecture Periods: -	Tutorial Periods: -	Practical Periods: 30	Total Periods: 30

COs/POs/PSOs Mapping

COs		Program Outcomes (POs)												Program Specific Outcomes (PSOs)			
	PO1	PO2 PO3 PO4 PO5 PO6 PO7 PO8 PO9 PO10 PO11 PO											PSO1	PSO2	PSO3		
1	3	2	2	2	-	-	-	-	3	3	-	1	1	1	1		
2	3	3	3	2	2	2	2	2	3	3	3	1	2	2	2		
3	3	3 2 2 1 - 2 - 3 3 3 1										1	2	2	2		

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

		Review 1			Review 2			Total
Assessment	Novelty	Presentation	Viva	Presentation	Demonstration	Viva	Report	Marks
Marks	10	20	10	20	20	10	10	100

2-18/-

Department	Artificial Intelligence and Data Science	Programme: B. Tech.							
Semester	VI	Course Category: AEC End Semester Exam Type:						pe: -	
Course Code	U23ADC6XX	Pe	eriods/W	eek	Credit	Max	imum Ma	arks	
Course Code	UZSADCOXX	L	Т	Р	С	CAM	ESE	TM	
Course Name	CERTIFICATION COURSE - VI	0	0	4	-	100	-	100	
	AI &	DS			<u> </u>		-	<u> </u>	
Prerequisite	-								

Students shall choose an International / Reputed organization certification course of 40-50 hours duration specified in the curriculum (It is mandatory to do a minimum of six courses) which will be offered through the Centre of Excellence. These courses have no credit and will not be considered for CGPA calculation.

- (i) Certification Courses are required to be completed to fulfil the degree requirements. All Certification courses are assessed internally for 100 marks.
- (ii) The Course coordinator handling the course will assess the student through attendance and MCQ test, and declare the student as "pass" on satisfactory completion. A letter grade "P" is awarded to declare pass.
- (iii) The marks scored in these courses will not be taken into consideration for the SGPA / CGPA calculations in the grade sheet.

Evaluation Methods

Accessment	Continuous Assessi	ment Marks (CAM)	Total Marka
Assessment	Attendance	MCQ Test	Total Marks
Marks	10	90	100



Department	Artific	Artificial Intelligence and Data Science Programme: B. Tech.									
Semester	VI		Cours	e Catego	ory: MC	End Sem	nester Ex	am Type): -		
Course Code	1123 V	DM606	Pe	eriods/W	eek	Credit	Max	imum Ma	arks		
Course Code	UZJA	Б ІМООО	L	Т	Р	С	CAM	ESE	TM		
Course Name	GENE	PER EQUALITY	2	0	0	-	100	-	100		
Prerequisite	-							BT N	/lapping		
	On co	empletion of the course, the studen	its will be	e able to)				est Level		
	CO1	CO1 Describe the general identity, social construction of gender roles.									
Course	CO2 Illustrate the causes and issues of gender discrimination in Indian society.										
Outcomes	CO3	CO3 Describe the workplace discrimination, media influences on gender and culture.									
	CO4 Familiarize with international and Indian frameworks on gender equality.								K2		
	CO5	Illustrate the current challenges in gender of technology.	er equality	, including	g the glas	ss ceiling a	nd the role	•	K2		
UNIT – I	Introduction to Gender Equality Periods:06										
OIIII — I							0.00				
Gender equality	/ – explo	ring gender identity and expression, Untives on gender roles, Analyzing key mile				truction of		oles and	CO1		
Gender equality	/ – explo	ring gender identity and expression, Un	stones in			truction of	general ro	oles and	CO1		
Gender equality norms, historica UNIT – II Gender discrimi social beliefs, pi	Gendential	ring gender identity and expression, Un ctives on gender roles, Analyzing key mile	estones in S quality – I	the fight for	or gender	truction of equality. Periods al set up, la	general ros	areness,	CO1		
Gender equality norms, historica UNIT – II Gender discrimi social beliefs, pi	Genderication in ractice and	ring gender identity and expression, Unctives on gender roles, Analyzing key mile er Inequality and Its Manifestations Indian society – causes of gender inected custom – Issues of gender discriminat	estones in S quality – I	the fight for	or gender	truction of equality. Periods al set up, la	general rose. s:06 ack of awark, poor ea	areness,			
Gender equality norms, historica UNIT – II Gender discrimi social beliefs, prand health, viole UNIT – III Workplace discrimination	Genderimination	oring gender identity and expression, Untives on gender roles, Analyzing key mile er Inequality and Its Manifestations Indian society – causes of gender ineed custom – Issues of gender discriminat exploitation in workplace.	estones in G quality – I ion – Child	the fight for	or gender patriarcha e, child d	Periods Periods	general rose. s:06 ack of awark, poor ecc. s:06	areness, ducation			
Gender equality norms, historica UNIT – II Gender discrimi social beliefs, prand health, viole UNIT – III Workplace discrimination	Genderication in Gender	oring gender identity and expression, Unctives on gender roles, Analyzing key mile er Inequality and Its Manifestations Indian society – causes of gender inected custom – Issues of gender discriminate exploitation in workplace. Er and Culture 1, Media influences on gender and culture	estones in G quality – I ion – Child	the fight for	or gender patriarcha e, child d	Periods Periods	general rose. s:06 ack of awark, poor ecc. s:06 ety. Strate	areness, ducation	CO2		
Gender equality norms, historica UNIT – II Gender discrimi social beliefs, properties and health, viole UNIT – III Workplace discriming promoting gender UNIT – IV Gender Equality	Genderination in ractice and Genderimination er equality and Hu	oring gender identity and expression, Untives on gender roles, Analyzing key mile er Inequality and Its Manifestations Indian society – causes of gender inected custom – Issues of gender discriminate exploitation in workplace. For and Culture In, Media influences on gender and culture by and cultural understanding.	estones in quality – I ion – Child re, Gende	lliteracy, pd marriage	patriarcha e, child d wer dynar	Periods Periods Periods Periods Periods Periods Periods Periods	general rose. s:06 ack of awark, poor ed s:06 ety. Strate s:06 Equality un	areness, ducation egies for	CO2		
Gender equality norms, historica UNIT – II Gender discrimi social beliefs, prand health, viole UNIT – III Workplace discriming gender UNIT – IV Gender Equality Indian Constitution	Genderination in Gender	oring gender identity and expression, Untives on gender roles, Analyzing key mile er Inequality and Its Manifestations Indian society – causes of gender inected custom – Issues of gender discriminate exploitation in workplace. Er and Culture In, Media influences on gender and culture yand cultural understanding. Toting Gender Equality man Rights – International frameworks a	quality – I quality – I ion – Child re, Gendel	lliteracy, pd marriager and powentions on ategies fo	patriarcha e, child d wer dynar	Periods Periods Periods Periods Periods Periods Periods Periods	general rose. s:06 ack of awark, poor ed s:06 ety. Strate s:06 Equality un Equality in	areness, ducation egies for	CO2		
Gender equality norms, historica UNIT – II Gender discrimi social beliefs, prand health, viole UNIT – III Workplace discriptor of the context of the con	Genderination in a content of the co	pring gender identity and expression, Unctives on gender roles, Analyzing key mile er Inequality and Its Manifestations Indian society – causes of gender inequality and custom – Issues of gender discriminate exploitation in workplace. Er and Culture In, Media influences on gender and culture and cultural understanding. Toting Gender Equality The man Rights – International frameworks a ficies and initiatives for gender mainstream	quality - I quality - I ion - Child re, Gender and Conve	lliteracy, pd marriager and powentions on ategies fo	patriarcha e, child d ver dynar Gender r promoti	Periods Periods Periods Periods Periods Periods Periods Periods Periods Periods Periods Yeriods Periods	general ro s:06 ack of awa rk, poor ec s:06 ety. Strate s:06 Equality un Equality in s:06	egies for	CO2		

Text Books

- 1. "Gender and Society" by Raewyn Connell This book provides a comprehensive overview of gender roles, power dynamics, and the social construction of gender.
- 2. "The Second Sex" by Simone de Beauvoir A historical and philosophical examination of women's oppression and gender inequality.
- 3. "Women and Gender in the Indian Society" by Neera Desai and Usha Thakkar Focuses on the context of gender roles, inequality, and feminist movements in India.

Reference Books

- 1. Woman in early Indian societies, New Delhi: Manohar Publications. Sita A. Raman (2009).
- 2. A social and Cultural history, Volume1. Connecticut: Oxford: Praeger. Sita Raman (2009).
- 3. A social and Cultural history, Volume2. Connecticut: Oxford: Praeger.
- 4. Iftikhar R. (2016). Indian Feminism: Class, Gender and Identity in Medieval Ages. Chennai: Notion Press. Iftikhar, R. (2012).

Web References

- 1. https://www.unwomen.org
- 2. https://ncw.nic.in
- 3. https://en.unesco.org/themes/gender-equality
- 4. https://www.weforum.org/reports
- 5. https://wcd.nic.in

2.13/-

COs/POs/PSOs Mapping

COs					Program Specific Outcomes (PSOs)										
	PO1	01 PO2 PO3 PO4 PO5 PO6 PO7 PO8 PO9 PO10 PO11 PO12										PO12	PSO1	PSO2	PSO3
1	1	-	-	-	-	-	ı	-	ı	3	-	1	1	-	1
2	1	-		-	-	-	-	-	-	3	-	1	1	-	1
3	1	-	-	-	-	-	-	-	-	3	-	1	1	-	1
4	1	-	-	-	-	-	ı	-	ı	3	-	1	1	-	1
5	1	-	-	-	-	-	ı	-	-	3	-	1	1	-	1

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

	Continuo	us Assessment	Marks (CAM)	
Assessment	Attendance	MCQ Test	Presentation / Activity / Assignment	Total Marks
Marks	10	30	60	100

2-13/-

Department	Artificial Intelligence and Data Science	Progran	nme: B	.Tech.				
Semester	V/VI	Course	Catego	ry Code	: OE *Er	nd Semest	ter Exam Ty	/pe: TE
0	LIONA DROOM	Perio	ds / W	eek	Credit	Ma	ximum Mar	ķs
Course Code		L	Т	Р	С	CAM	ESE	TM
Course Name	PRINCIPLE OF ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING	3	0	0	3	25	75	100
	(Common to EEE, ECE, CSE, IT, ICE	, MECH,	CIVIL,	CCE, BI	ME, Mechati	ronics)		
Prerequisite	NIL							
	On completion of the course, the stude	nts will b	e able	to			BT Ma (Highest	
	CO1 Interpret foundational principles of artific	ial intellige	nce				K2	2
	CO2 Identify formal methods of knowledge re	presentation	on				K	2
Course	CO3 Interpret the fundamental issues and ch	allenges of	Reasor	ning			K	2
Outcomes	CO4 Apply mathematical relationships with Napplications	/lachine Le	arning a	algorithms	s for real worl	d	K	3
	CO5 Identify various deep learning and reinfo	rcement le	arning t	echnique	s to improve	the	K	2
Unit- I	Introduction				Periods: 0	_	•	
	rtificial Intelligence - Artificial Intelligence Problematation - Branches of Artificial Intelligence - Applica					duction Sys	stems - State	CO1
Unit- II	Knowledge Representation				Periods: 0	9		
	agement - Types of Knowledge - Knowledge repre esentation - Knowledge base. First order Logic – F					epresentatio	on - Issues ir	CO2
Unit- III	Reasoning				Periods: 0	9		
Types of reason	ng - reasoning with Fuzzy Logic - Rule based Reas	soning - Dia	agnosis	Reasonin	g.			CO3
Unit- IV	Learning				Periods: 0	9		
	ng - Machine Learning - Intelligent agents - As CADA Application – k-Means Clustering - Fuzzy (Case Stud	ly: Custome	CO4
Unit- V	Reinforcement and Statistical Learning				Periods: 0	9		.i
	Problem - Hidden Markov Model - Linear Classif rocess - Types of Network – Perceptron - RBF Ne						work – ANN	CO5
Lecture Perio	ds: 45 Tutorial Periods:	Practic	al Perio	ods: -	Т	otal Perio	ods: 45	
2014. 2. Tom M	Hareendran S., Anand Hareendran, And Vinod Ch Mitchell, "Machine Learning", McGraw-Hill Sciend Jarrington, "Machine Learning in action", Manning	ce, 1997.		al Intellige	ence and Mac	hine Learn	ing" PHI Pub	lication,
Reference Bo			, _ • · _ ·					
1. Charu Discov 2. Andrea 3. Eremy	C. Aggarwal "Data Classification Algorithms an ery Series. s C. Mueller and Sarah Guido, "Introduction to Ma Watt, Reza Borhani, and Aggelos K. Katsaggelos dge University Press, 2016.	chine Lear	ning wit	h Python'	", O'Reilly Me	edia, Inc. Fi	rst Edition, 2	016.
	nalev-Shwartz and Shai Ben-David, "Understandin	g Machine	Learnir	g: From	Theory to Alg	orithms", C	ambridge U	niversity

Web References

- https://www.coursera.org/learn/machine-learning
 https://ml-cheatsheet.readthedocs.io/en/latest/regression_algos.html
 https://machinelearningmastery.com/a-tour-of-machine-learning-algorithms

COs/POs/PSOs Mapping

COs		Program Outcomes (POs)													ecific (SOs)
	PO1	PO1 PO2 PO3 PO4 PO5 PO6 PO7 PO8 PO9 PO10 PO11 PO1											PSO1	PSO2	PSO3
1	2	2	2	_	1	-	-	-	-	-	-	-	2	2	2
2	1	2	2	_	_	-	-	-	-	-	-	-	1	1	-
3	2	2	1	2	_	-	-	-	-	-	-	-	-	1	1
4	3	2	2	2	1	-	-	-	-	-	-	-	1	-	1
5	2	2	2	2	1	ı	-	-	-	-	i	-	1	1	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assassment		End Semester	Total				
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	5	5	5	75	100

^{*}Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

Department	Artific	ial Intell	igence and	l Data Science	Pro	gram	me: B	в.Т	ech.					
Semester	V/VI				Cou	ırse (Categ	ory	Co.	le: O	E *E	nd Semes	ter Exam T	ype: TE
					F	Perio	ds / W	ee	k		Credit	Ma	ximum Ma	rks
Course Code	U23AI	DOC01			L	_	Т		Р		С	CAM	ESE	TM
Course Name	INTRO	DUCTIO	N TO DAT	A SCIENCE	3	3	0		0		3	25	75	100
	(Common	to EEE, E0	CE, CSE, IT, IC	E, ME	CH, (CIVIL,	C	CE,	ЗМЕ,	Mecha	tronics)		
Prerequisite	NIL													
	On co	ompletio	n of the co	urse, the stud	lents w	ill be	able	to)				BT Ma (Highes	
	CO1	Explore t	he fundamen	ital concepts of d	lata scie	ence.							K	2
	CO2	Interpret	the Mathema	atical Knowledge	for Data	a Scie	ence re	qu	ires	o ma	nipulate	the given da	ata K	2
Course	CO3	Visualize	and present	the inference us	ing vari	ous to	ools.						K	3
Outcomes	CO4	Identify d	ifferent oppo	rtunities in Indus	tries an	d real	lize it ir	n ap	pplic	ations			K2,	K3
	CO5	Interpret	the ethics ne	eded for maintai	ning pri	vacy,	data s	har	ring a	nd de	ecision-n	naking.	K	2
Unit- I	Introd	uction to	Data Scie	nce						Pe	eriods:	09	<u> </u>	
Definition – Big Da Scientist - Data So – Presentation.	ata and	Data Scie	nce Hype – \	Why data science										
Unit- II	Math	ematical	Preliminar	ies						Pe	eriods:	09		
Probability: Proba Descriptive Statist Analysis: Correlat	tics: Cer	ntrality Me	asures – Var	iability Measures	s - Interp	oretin	g Varia	nc	e – C					
Unit- III	<u> </u>	Science		<u>g</u>						Pe	eriods:	09		
Introduction to D	ata Scie	nce Tool -	– Data Clean	ing Tools – Data	Mungir	ng an	d Mod	elli	ing T	ools –	- Data Vi	sualization	Tools – Too	ls
for Data Science.	7									-				CO3
Unit- IV				tunities and A						<u>i</u>	eriods:			
Data Economy a Introduction, Gen- and Humanities - Engineering – Fin	eral App - Bioinfo	lication Gormatics -	uidance - Dit - Consulting	fferent Domain – Services – Ecc	- Advert	ising	– Aero	spa	ace a	nd A	stronomy	/ – Arts, Cre	eative Desig	n co4
Unit- V	Ethic	s and Re	ecent Trend	sk						Pe	eriods:	09		
Data Science Eth consent - The Five					he data	ı - Va	lluing o	diffe	erent	aspe	cts of pr	ivacy - Get	ting informe	cO5
Lecture Period	s: 45		Tutorial F	eriods:	Pra	ctica	al Peri	od	ls: -		'	Total Peri	ods: 45	
Text Books														
 Chirag S SinanOz 	hah, "A demir, "l	Hands on Principles	Introduction of Data Scie	hamed Ali, "Intro to Data Science nce", Packt Publ , "Ethics and Dat	", Camb lication,	ridge 2016	Unive	rsit	ty Pre	ess, 2	020.	ons Co., 1st	edition, 201	6.
Reference Boo	······	*145011, IVII	ino Lounides,	Lunes and Dat	u 0010111	JU , C		, ,!	. J. U		2010.			
1. Hector G 2. Paul Cur 3. Steven S 4. Rajendra 5. Longbing	suerrero zon, Pe S. Skiena Akerka J. Cao "	ter W. Mc a, "Data S ır, Priti Sri Data Scie	Owan, "The cience Desig nivas Sajja, "	Modeling and S Power of Compu in Manual", Sprin Intelligence Tech g: The Next So	utational ng Intern nniques	l Thin nation for D	king", \ al Pub ata Sc	Wo lica ien	orld S ation, ce",	cienti 2017 Sprinç	fic Publis g Interna	shing, 2017. tional Public	cation, 2016	
Publicati		3.												
2. https://w	ww.yout ww.java	tpoint.com	n/data-scienc		_channe	el=ed	ureka%	621						
3. https://w	ww.cour	sera.org/b	orowse/data-	science /										

COs/POs/PSOs Mapping

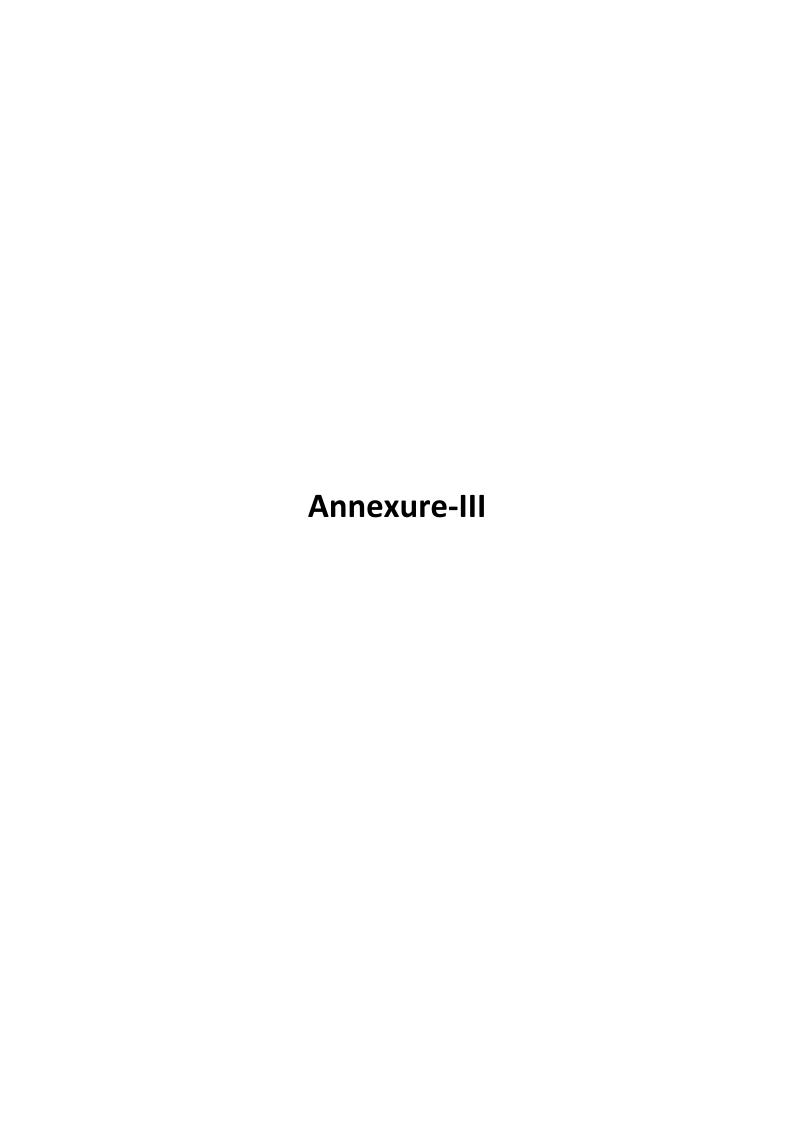
COs					Prog	ram O	utcom	es (PO	s)				Program Specific Outcomes (PSOs)			
	PO1	PO2	PO3	PO4	PO5	P06	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3	
1	1	2	2	2	2	-	-	-	-	-	-	-	-	-	-	
2	2	2	2	1	1	1	1	1	1	-	ı	-	i	-	-	
3	2	1	2	2	1	ı	ı	ı	ı	į	ı	ı	ı	1	-	
4	1	2	2	1	1	ı	1	1	ı	ı	ı	ı	į	-	-	
5	2	1	1	2	1	ı	-	1	-	1	1	1	1	1	-	

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Accoment		Continuou	s Assessmer	nt Marks (CAM)		End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	5	5	5	75	100

^{*}Application oriented / Problem solving / Design / Analytical in content beyond the syllabus



Department												
Semester	IV		Course	Catego	ry: PC	*End	Seme	mester Exam Type: T Maximum Marks				
Course Code	U23AI	DV401	Perio	ods / We	ek	Credit	١	/laximum l	Marks			
Course Code	UZSAI	DA401	L	Т	Р	С	CAM	ESE	Ξ	TM		
Course Name		LLEL PROGRAMMING AND HIGH ORMANCE COMPUTING	3	1	0	4	25	75	•	100		
-		Common to all I	Branches	except	AI & DS	3	<u>+</u>					
Prerequisite	Basic	s of Programming (C, C++), Linux O	perating	System	S							
	On co	ompletion of the course, the stude	nts will l	oe able	to			BT Map Level)	ping (Hi	ghest		
	Apply MPI framework for passing message parallelly across processes								K3			
Course Outcomes	CO2	Apply Pthreads for creating shared me	mory para	allel prog	rams				K3			
Outcomes	CO3	Apply OpenMP paradigms to create sh	nared men	nory para	llel prog	rams		K3				
	CO4	Apply either OpenMP, MPI for parallel	algorithms	for sear	ching an	d sorting			K3			
	CO5	Apply CUDA programming for configurand CPU	ring hard	ware and	l transfe	r data across	GPU		K3			
UNIT-I	Mess	sage Passing Paradigm			Р	eriods: 12	L					
		v – Message matching – MPI I/O – Pa t, scatter, gather, allgather – Derived typ							CC)1		
UNIT-II	Share	ed Memory Paradigm: Pthreads			Р	eriods: 12						
		read synchronization – Critical sections d write locks with examples - Caches, ca							C	O2		
UNIT-III	Share	ed Memory Paradigm: OpenMP			Р	eriods: 12		<u></u>				
	nization	 s – scope of variables – Reduction clause in OpenMP – Case Study: Producer-C practices 				-		- :	CC)3		
UNIT-IV	··•	llel Algorithms			Р	eriods: 12						
• •	•	ithms: Reduction – Broadcast - Prefix sissors and multicomputer. Sorting: Odd						•	CC)4		
UNIT-V	GPU	Programming with CUDA			P	eriods: 12		<u>i</u>				
GPUs and GPGF	PU - GPU - CUDA	J architectures - Heterogeneous comput trapezoidal rule – improvements - Imple						•	CC)5		
Lecture Period	ls: 45	Tutorial Periods: 15	Pra	ctical P	eriods:	-		Total Per	iods: 60)		

Text Books

- Peter S. Pacheco, Matthew Malensek, "An introduction to parallel programming", Second edition, Morgan Kaufmann, 2021
- Niranjan N. Chiplunkar, Raju K, "Introduction to Parallel Computing", Wiley, 2021.
- 3. Michael J. Quinn, "Parallel Computing: Theory & Practice", Tata McGraw Hill, Second edition, Reprint 2017.

Reference Books

- A. Munshi, B. Gaster, T. G. Mattson, J. Fung, and D. Ginsburg, "OpenCL programming guide", Addison Wesley, 2011
 M. J. Quinn, "Parallel programming in C with MPI and OpenMP", Tata McGraw Hill, 2011.
- 3. Rob Farber, "CUDA application design and development", Morgan Haufmann, 2011

Web References

- http://condor.cc.ku.edu/~grobe/docs/intro-MPI-C.shtml
- http://www.hpcc.unn.ru/mskurs/ENG/DOC/pp09.pdf
- 3. https://www.cs.cmu.edu/afs/cs/academic/class/15492-f07/www/pthreads.html
- https://www.openmp.org/ 4.
- 5. https://developer.nvidia.com/blog/even-easier-introduction-cuda/

5- NS/-

POs/PSOs Mapping

COs	Progr	am Ou	tcome	s (POs)								Program Specific Outcomes (PSOs)		
000	PO1	PO2	PO3	PO4	PO5	PO6	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	3	3	3	3	-	-	-	-	-	-	-	3	3	2
2	3	3	3	3	1	-	-	-	-	-	-	-	3	3	2
3	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3
4	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3
5	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

		Contir	nuous Ass	essment Marks	(CAM)	End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	5	5	5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

2-18/-

Department	Artific	cial Intelligence and Data Science	Progran	nme: B	.Tech / (Honour / M	linor)		
Semester	V		Course	Catego	ry: PC	*End	d Semeste	r Exam Typ	oe: TE
Course Code	11224	DVE00	Perio	ods / Wo	eek	Credit	Max	kimum Mar	ks
Course Code	UZSA	DX502	L	Т	Р	С	CAM	ESE	TM
Course Name	ADVA	NCED DEEP LEARNING	3	1	0	4	25	75	100
	.,	Common to all Branch	ies excep	t Al & [DS				
Prerequisite	1	nderstanding of basic machine learni Programming	ng conce	pts and	l neural r	network arc	hitectures,	Proficienc	y in
	On co	ompletion of the course, the studer	nts will b	e able	to			1	apping st Level
Course	CO1	Apply different Convolutional Neural No	etwork for	process	ing image	es effectively	'	·····	(3
Outcomes	CO2	Apply generative models for image syn	nthesis, sty	le trans	fer, and d	lata augment	ation tasks.	·	(3
	CO3	Apply advanced sequence models for multimodal learning.	NLP task	s like te	ext genera	ation, summa	arization, ar	nd k	(3
	CO4	Apply reinforcement learning concepts	in AI, rob	otics, an	d autonoi	mous system	ıs.	ŀ	(3
	CO5	Apply techniques such as neural archit for edge devices and energy-efficient A			d meta-lea	arning to opt	imize mode	ls ř	(3
UNIT – I	Advai	nced Convolutional Networks and	Visual R	ecogni	tion	Periods:12	2	-	
tyleGAN). Variat	Gener sarial Ne sional Au	rative Models tworks (GANs): Architecture, loss function utoencoders (VAEs): Probabilistic interp	oretation,	KL dive	ergence,	reconstruction	nts (DCGA on. Diffusio	n Models:	CO2
ata augmentation	l <u>.</u>	bilistic models for image generation. Appl					-	olution, and	
UNIT – III		ence Models and Natural Language				Periods:12			T
elf-attention med	hanism, ng: Com	s (RNNs): Advanced concepts in LSTMs positional encoding. BERT, GPT, T5 No bining text, images, and audio for riche analysis.	Models: P	re-traini	ng, fine-t	uning, mask	ed languag	ge models.	CO3
UNIT – IV	Reinf	orcement Learning and Deep RL				Periods:12	2		
earning with deep	networl	ent Learning: Markov decision processes ks, experience replay. Policy Gradient N RL Applications: Game AI, robotics, autor	Methods: I	REINFO	RCE, act	tor-critic met	hods, Proxi		CO4
UNIT – V	···	al Architecture Search and Model C				Periods:12	_		
Quantization. Kn	owledge	ch (NAS): Techniques for automating architecture. Distillation: Transfer of knowledge from tasks with limited data. Applications: Op	large mo	dels to	smaller, I	Meta-Learnir	ng: Few-sho	ot learning,	CO5
Lecture Period	ls:45	Tutorial Periods: 15	Practic	al Perio	ods: -	Т	otal Perio	ds:60	.L
extbooks						<u>i</u>			
2. Charu C.	Aggarwa	oshua Bengio, and Aaron Courville, "Dee al, "Neural Networks and Deep Learning", Transformers for Natural Language Proce	Springer,	2018.					

- 3. Denis Rothman, "Transformers for Natural Language Processing", Packt Publishing, 1st Edition, 2021.
- 4. Frank Hutter, Lars Kotthoff, and Joaquin Vanschoren, "AutoML: Methods, Systems, Challenges", Springer Cham, 1st Edition, 2021.

References

- 1. Cosma Rohilla Shalizi, "Advanced Data Analysis from an Elementary Point of View", Cambridge University Press, 2015.
- Deng & Yu, "Deep Learning: Methods and Applications", Now Publishers, 2014.
- Michael Nielsen, "Neural Networks and Deep Learning", Determination Press, 2015.
- Josh Patterson, Adam Gibson, "Deep Learning A Practitioner's Approach", O'Reilly Media, 2017.
- 5. Nikhil Buduma, "Fundamentals of Deep Learning", O'Reilly, 2017.

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Web References

- https://nptel.ac.in/courses/106/106/106106184/
 http://deeplearning.net/Dj
- http://deeplearning.net/Dj
 https://www.guru99.com/deep-learning-tutorial.html
 https://www.coursera.org/specializations/deep-learning
- 5. http://neuralnetworksanddeeplearning.com

COs/POs/PSOs Mapping

COs	Progra	am Out	tcome	s (POs)								Program Specific Outcomes (PSOs)		
	PO1	PO2	PO3	PO4	PO5	PO6	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	3	3	3	3	-	-	-	-	-	-	-	3	3	2
2	3	3	3	3	1	-	-	-	-	-	-	-	3	3	2
3	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3
4	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3
5	3	3	3	3	1	-	-	-	-	_	-	-	3	3	3

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

		Contin	uous Ass	essment Marks	(CAM)	End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	,	10	5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

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Semester	VI				Course	Catego	ry: PC	*En	d Seme	ster Exam Ty	pe: TE	
Course Code	1100 A	DVc03			Perio	ds / We	eek	Credit	N	Maximum Ma CAM ESE 25 75 S BT Mappir (Highest Lev K2 K3 n. K3 n. K3 rohom Variables, oblem Solving — for Episodic and ronous dynamic ion, Monte Carlo ing Starts.	rks	
Course Code	UZSA	DX603			L	Т	Р	С	CAM	ESE	TM	
Course Name	REIN	FORCEME	NT LEARN	ING	3	1	0	4	25	75	100	
			Co	ommon to all E	Branches (except A	AI & DS		<u>i</u>			
Prerequisite	Mach	ine Learnir	ng, Program	ming in Pytho	n, knowle	dge of F	Probabili	ty and stati	stics			
	On c	ompletion	of the cou	rse, the stude	ents will b	e able	to			BT Mappin (Highest Lev	_	
	CO1 Interpret the concepts of Reinforcement Learning to solve real world problems.									K2		
Course Outcomes	CO2	1	kov Decision uation and pr	Process, Monte	Carlo, Ter	nporal D	ifference	methods fo	r	. K2 K3 . K3		
	CO3	Apply the	•••••••••••••••••••••••••••••••••••••••	Methods and	l On-poli	cy Pre	ediction w	ith Approxim	nation.	К3		
	CO4	Apply suita	able Reinforce	ement Technique	es for a giv	en probl	em			К3		
	CO5			ces, REINFORC arlo tree search						КЗ		
UNIT-I	Rein	<u>.</u>	•••••		111101111010	CITICITE						
			t Learning	Primitives			. P	eriods: 12				
Introduction and	-			Primitives Framework, Pro	obability	Basics:	<u>L</u>		. Randor	n Variables.	CO1	
	d Basics	of RL, Def	fining RL	Framework, Pro	•		Probab	ility Axioms		· 1	CO1	
robability Mass	d Basics Function	of RL, Def on, Probabi	fining RL	Framework, Pro	•		Probab	ility Axioms		· 1	CO1	
Probability Mass Searching, Logica	d Basics Functional Agent	of RL, Def on, Probabi s.	fining RL	Framework, Pro Function, Introd	•		Probab Intelliger	ility Axioms		· 1	CO1	
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- 1. Russell, S. J., & Norvig, P. "Artificial Intelligence: A Modern Approach", 4th Edition, Pearson, 2020.
- 2. Busoniu, L., Babuška, R., & De Schutter, B., "Reinforcement Learning and Dynamic Programming Using Function Approximators". CRC Press, 2010
- 3. Richard S. Sutton, Andrew G. Barto, "Reinforcement Learning, An Introduction: Adaptive Computation and Machine Learning series", MIT Press, 4th edition, 2018
- 4. Paul A. Gagniuc, "Markov Chains: From Theory to Implementation and Experimentation", John Wiley & Sons, 2017, ISBN 1119387558, 9781119387558



Reference Books

- 1. Sutton, R. S., & Barto, A. G. "Reinforcement Learning: An Introduction", 2nd Edition, 2018.
- 2. Silver, D."UCL Course on Reinforcement Learning", 2015..
- 3. Mnih, V., et al., "Human-level control through deep reinforcement learning". Springer Nature, 2015.

Web References

- 1. https://www.datacamp.com/tutorial/reinforcement-learning-python-introduction
- 2. https://medium.com/analytics-vidhya/a-beginners-guide-to-reinforcement-learning-and-its-basic-implementation-from-scratch-2c0b5444cc49
- 3. https://towardsdatascience.com/reinforcement-learning-101-e24b50e1d292
- 4. https://towardsdatascience.com/introduction-to-reinforcement-learning-rl-part-5-monte-carlo-methods-25067003bb0f

COs/POs/PSOs Mapping

COs	Progra	am Ou	tcome	s (POs)								Program Specific Outcomes (PSOs)		
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	3	3	3	3	-	-	-	-	-	-	-	3	3	2
2	3	3	3	3	1	-	-	-	-	-	-	-	3	3	2
3	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3
4	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3
5	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Evaluation Me	tinous						
		Contin	uous Ass	essment Marks	(CAM)	End Semester	Total
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	5	5	5	75	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

2-13/-

Semester		J.u	ligence and Data Science				Honour / M				
Semester	VII			Course	Catego	ry: PC	*End	Semest	ter Exam T	ype: TE	
Caa. Ca.da	1100 A	DV704		Perio	ods / We	eek	Credit	М	aximum M	arks	
Course Code	U23A	DX704		L	Т	Р	С	CAM	ESE	TΝ	
Course Name	IMAG	E AND \	/IDEO ANALYTICS	3	1	0	4	25	75	100	
			Common to all E	Branches	except /	AI & DS					
Prerequisite	Basic i	mage pro	ocessing concepts, knowled	ge of Mad	chine Le	earning a	nd Deep Le	arning f	undamenta	als	
	On c	ompletion	on of the course, the stude	ents will b	e able	to		(BT Mappi Highest Le		
Course	CO1	video file		_					K	2	
Outcomes	mes image analysis K2										
	CO3		arious concepts of Image analy			-			K	3	
	CO4	Apply r	multifarious feature detection ar	nd descript	ion techi	niques for	image analy	rtics	K	3	
	CO5	Apply v	rarious object detection and rec	ognition te	chnique	s for track	ting objects in	า	K	3	
UNIT-I	Intro		to Digital Image and Video	Process	sing	Pe	eriods: 12	<u>i</u>			
-	Video T	ransmiss									
	Operati	ions. Bina	ion. Gray-Level Processing: Imry Image Processing, Binary Im	nage Morp	hology.		-	etween In	nages -		
UNIT-II	Operati	ions. Bina je and V	ry Image Processing, Binary Imideo Enhancement and Re	nage Morp	hology. 1	Po	eriods: 12				
UNIT-II patial domain - L letection and Re correction - Flick	Imag Imag inear ar moval - ker Para	ions. Bina ge and V nd Non-lin Blotch Do	ry Image Processing, Binary Im	nage Morplestoration ering - Fred and Interp	hology.	P omain– H	eriods: 12 omomorphic Intensities -	Filtering Intensity	- Blotch Flicker	CO2	
UNIT-II patial domain - L etection and Re correction – Flick econvolution filte	Operati Imag inear ar emoval - ker Para ers.	ions. Bina ge and V nd Non-lin Blotch Do nmeter Es	ry Image Processing, Binary Imideo Enhancement and Re ear Filtering - Morphological filte etection - Motion Vector Repair	nage Morplestoration ering - Fred and Interp	hology.	Pomain– H Corrupted c method	eriods: 12 omomorphic Intensities -	Filtering Intensity	- Blotch Flicker	CO2	
UNIT-II patial domain - L petection and Re correction – Flick econvolution filte UNIT-III mage Compress ompression. Vid	Image concentration of the con	ions. Bina ge and V nd Non-lin Blotch De meter Es and Vic ffman coe pression:	ideo Enhancement and Re ear Filtering - Morphological filte etection - Motion Vector Repair timation - Wavelet based imag	estoration estoration ering - Fred and Interprete denoising	hology.	Pomain— H Corrupted c method Pomess Codi	eriods: 12 omomorphic I Intensities - s for image r eriods: 12 ng - Wavele	Filtering Intensity estoration	- Blotch Flicker n using	CO2	
UNIT-II Spatial domain - L Detection and Re Correction - Flich econvolution filte UNIT-III mage Compress	Image in the control of the control	ions. Bina ge and V nd Non-lin Blotch Do ameter Es and Vio ffman coo pression: s.	ry Image Processing, Binary Imideo Enhancement and Ree ear Filtering - Morphological filterection - Motion Vector Repair timation - Wavelet based imagedeo Analysis ding - Run length coding - La	estoration estoration ering - Fred and Interprete denoising	hology.	Pomain— H Corrupted c method Poess Codi and the H	eriods: 12 omomorphic I Intensities - s for image r eriods: 12 ng - Wavele	Filtering Intensity estoration	- Blotch Flicker n using		
UNIT-II patial domain - L etection and Re correction – Flick econvolution filte UNIT-III mage Compress compression. Vid IPEG-2 Video S UNIT-IV htroduction to fea	Image in an are moval - ker Para ers. Image ision: Hule o Company tandard: Feature de ature de ature de ature de	ions. Bina ge and V nd Non-lin Blotch Do ameter Es and Vic ffman coo pression: s. re Detectors - c egion Bas	ideo Enhancement and Ree ear Filtering - Morphological filtetection - Motion Vector Repair timation - Wavelet based imagedeo Analysis ding - Run length coding - La Basic Concepts and Technique	estoration estoration ering - Fred and Interple e denoisin ZW coding es of Video king - Basio owing and	hology. n quency depolating (g - Basic - Lossl Coding a c edge d Region	Pomain—H Corrupted comethod Pomess Codi and the H Pometectors—	eriods: 12 omomorphic Intensities - s for image r eriods: 12 ng - Wavele .264 Standar eriods: 12 canny - sob	Filtering Intensity estoration ts based d - MPEC	- Blotch Flicker n using image G-1 and		
UNIT-II patial domain - L etection and Re correction – Flick econvolution filte UNIT-III mage Compress compression. Vid IPEG-2 Video S UNIT-IV ntroduction to fea mage Segmenta asic global thres	Image in a company	ions. Bina ge and V nd Non-lin Blotch Do ameter Es e and Vio ffman coo pression: s. re Detect egion Bas - optimur	ideo Enhancement and Ree ear Filtering - Morphological filtetection - Motion Vector Repair timation - Wavelet based imaged eo Analysis ding - Run length coding - La Basic Concepts and Technique etion and Description descriptors - matching and tracked Segmentation - Region Gr	estoration estoration ering - Fred and Interple e denoisin ZW coding es of Video king - Basio owing and	hology. n quency depolating (g - Basic - Lossl Coding a c edge d Region	Properties of the properties of the Holland the Holland splitting and the splitting and the splitting and the splitting and the splitting and the splitting and the splitting and the splitting and the splitting and the splitting and the splitting are splitting as the splitting and the splitting are splitting as the splitting are splitting a	eriods: 12 omomorphic Intensities - s for image r eriods: 12 ng - Wavele .264 Standar eriods: 12 canny - sob	Filtering Intensity estoration ts based d - MPEC	- Blotch Flicker n using image G-1 and	CO3	
UNIT-II patial domain - L etection and Re correction – Flick econvolution filte UNIT-III mage Compress compression. Vid IPEG-2 Video S UNIT-IV ntroduction to fea mage Segmenta asic global thres UNIT-V Dbject detection	Image in the company of the company	ge and V nd Non-lin Blotch Do meter Es and Vic ffman coo pression: s. re Detec tectors - c egion Bas - optimur ect Detec ognition in	ideo Enhancement and Re ear Filtering - Morphological filte etection - Motion Vector Repair timation - Wavelet based imag deo Analysis ding - Run length coding - La Basic Concepts and Technique etion and Description descriptors - matching and track sed Segmentation — Region Gr in global thresholding using Ots etion and Recognition in image and video - basic textu	estoration ering - Fred and Interprete denoising ZW coding es of Video king - Basic owing and u's Method	hology. n quency depolating (g - Basic - Lossl Coding a c edge d Region	Pomain—H Corrupted comethod Pomess Codi and the H Pometectors— Splitting a	eriods: 12 omomorphic Intensities - s for image r eriods: 12 ng - Wavele .264 Standar eriods: 12 canny - sob and Merging eriods: 12	Filtering Intensity estoration ts based d - MPEC el - prew - Thresh	- Blotch Flicker n using image G-1 and itt etc olding—	CO3	
UNIT-II patial domain - L retection and Re correction – Flick econvolution filte UNIT-III mage Compress compression. Vid IPEG-2 Video S UNIT-IV ntroduction to fea mage Segmenta asic global thres UNIT-V	Image in a company	ge and V nd Non-lin Blotch Do meter Es and Vic ffman coo pression: s. re Detec tectors - c egion Bas - optimur ect Detec ognition in	ideo Enhancement and Re ear Filtering - Morphological filte etection - Motion Vector Repair timation - Wavelet based imag deo Analysis ding - Run length coding - La Basic Concepts and Technique etion and Description descriptors - matching and track sed Segmentation — Region Gr in global thresholding using Ots etion and Recognition in image and video - basic textu	estoration ering - Fred and Interprete denoising ZW coding es of Video wing - Basicowing and u's Method	hology. n quency depolating (g - Basic - Lossl Coding a c edge d Region	Pomain—H Corrupted comethod Pomethod ess Codinand the H Pometectors— Splitting a	eriods: 12 omomorphic Intensities - s for image r eriods: 12 ng - Wavele .264 Standar eriods: 12 canny - sob and Merging eriods: 12 and its appli	Filtering Intensity estoration ts based d - MPEC	- Blotch Flicker n using image G-1 and itt etc olding—	CO3	

- 1. Alan Bovik, "Handbook of Image and Video Processing", 2nd Edition, Academic Press, 2005.
- 2. Rafael C. Gonzalez and Richard E. Woods, "Digital Image Processing", 3rd Edition, Pearson Education, 2008.
- 3. Richard Szeliski, "Computer Vision Algorithms and Applications", Springer, 2011.
- 4. Ali Ismail Awad and Mahmoud Hassaballah, "Image Feature Detectors and Descriptors", Foundations and Applications, Springer; 1st ed. 2016 edition
- 5. Xiaoyue Jiang and Abdenour Hadid, "Deep Learning in Object Detection and Recognition Hardcover", Springer; 1st ed. 2019 edition (27 November 2019).

2-13/-

Reference Books

- 1. Anil K Jain, "Fundamentals of Digital Image Processing", PHI, 2011.
- 2. Oge Marques, "Practical Image and Video Processing Using MatLab ", Wiley, 2011.
- 3. John W. Woods," Multidimensional Signal, Image, Video Processing and Coding ", Academic Press, 2006.
- 4. Mohammed Salemdeeb, "Object Detection and Recognition Using Deep Learning", Scholars' Press, 2020.
- 5. Davut Armagan Kaya, "Feature Detection and Matching", Grin Verlag, 1st edition, 2021).

Web References

- 1. https://www.geeksforgeeks.org/digital-image-processing-basics/
- 2. https://www.javatpoint.com/digital-image-processing-tutorial
- 3. https://www.tutorialspoint.com/dip/index.htm

POs/PSOs Mapping

COs	Progra	ogram Outcomes (POs)												ram Spe omes (P	
	PO1	PO2	PO3	PO4	PO5	PO6	P07	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	3	3	3	3	-	-	-	-	-	-	-	3	3	2
2	3	3	3	3	1	-	-	-	-	-	-	-	3	3	2
3	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3
4	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3
5	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

		Contin	uous Ass	(CAM)	End Semester	Total	
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	10 5 5		5	75	100	

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

5.15/-

Department	Artificial Intelligence and Data Science Programme: B.Tech / (Honour / Minor)										
Semester	VIII		Course	Catego	ry: PC	*Enc	d Semester	Exam T	ype: TE		
Cauraa Cada	1100 A	DVOOF	Peri	ods / W	eek	Credit	Maxii	mum Ma	arks		
Course Code	UZ3A	DX805	L	Т	Р	С	CAM	ESE	TM		
Course Name	PRON	IPT ENGINEERING	3	1	0	4	25	75	100		
		Common to all B	ranches	except	AI & DS						
	Strong technic	understanding of Natural Language ques.	Processi	ng, fam	iliarity wi	th Machine	Learning m	odels aı	nd		
		ompletion of the course, the studer	nts will b	e able	to				Mapping nest Level)		
Course	CO1	Interpret the basic concepts and impor of prompts and their applications.	tance of p	rompt e	ngineerin	g, including v	arious types		K2		
Outcomes	CO2										
	CO3	Apply various metrics and feedback, a	nd improv	e promp	t effective	eness.			K3		
	CO4	Apply advanced techniques in prompt e	engineerir	ng for ha	ndling diff	erent types o	of prompts		K3		
	CO5	Apply prompt engineering concepts in	various fo	r buildin	g interact	ive systems.		К3			
UNIT-I	Intro	duction to Prompt Engineering			Р	eriods: 12					
	•	e and Applications – Types of Prompts in Prompt Engineering.	- Compo	onents o	f Effective	e Prompts –	Challenges	and	CO1		
UNIT-II	· . · · · · · · · · · · · · · · · · · ·	ning Effective Prompts			Р	eriods: 12		ii			
		ign – Structuring Prompts – Contextual R alls and How to Avoid Them.	Relevance	Clarity	and Pre	cision – Exar	mples and Be	est	CO2		
UNIT-III	Evalu	lating Prompt Performance			Р	eriods: 12		<u>i</u>			
	-	ctiveness – User Feedback and Iteratio Prompt Responsiveness – Tools for Eva		-		n Methods –	Analyzing U	Jser	CO3		
UNIT-IV	Adva	nced Techniques in Prompt Engin	eering		Р	eriods: 12		<u>.</u>			
	-	hniques – Leveraging Machine Learnii alization and Customization – Integrating	-	-	-		-		CO4		
UNIT-V	Case	Studies and Applications			Р	eriods: 12					
nteractive Syste	ms with	t Engineering Applications – Healthcare Prompts – Real-world Case Studies and igning a Prompt System.						- :	CO5		
LecturePeriod		TutorialPeriods: 15	Pra	cticalP	eriods: ·	•	Total Peri	ods:60			

Text Books

- John D. Kelleher, Brian Mac Namee, and Aoife D'Arcy, "Fundamentals of Machine Learning for Predictive Data Analytics: Algorithms, Worked Examples, and Case Studies," MIT Press, 2015.
 Christopher Manning, Hinrich Schütze, and Prabhakar Raghavan, "Introduction to Information Retrieval," Cambridge University Press, 2008.

- Kathleen R. McKeown, "Introduction to Natural Language Processing," McGraw-Hill, 1992.
 Jacob Andreas, "Task-Oriented Dialogue Systems for Conversational AI," Springer, 2020.

Reference Books

- 1. Daniel Jurafsky and James H. Martin, "Speech and Language Processing," 3rd Edition, Pearson, 2021.
- 2. Yoav Goldberg, "Neural Network Methods for Natural Language Processing," Morgan & Claypool Publishers, 2017.
- 3. Christopher D. Manning and Hinrich Schütze, "Foundations of Statistical Natural Language Processing," MIT Press, 1999.
- 4. Mike Lewis and Tom Kwiatkowski, "Advanced Methods for Natural Language Processing," Springer, 2022.

Web References

- 1. https://www.nltk.org/book/
- 2. https://github.com/dennybritz/deeplearning-pytorch
- 3. https://towardsdatascience.com/prompt-engineering-7e1666f71e7f
- 4. https://github.com/f/awesome-chatgpt-prompts
- 5. https://ai.googleblog.com/search/label/Dialog%20Systems

COs/POs/PSOs Mapping

COs	Progra	ogram Outcomes (POs)											Program Specific Outcomes (PSOs)		
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2	PSO3
1	3	3	3	3	3	-	-	-	-	-	-	-	3	3	2
2	3	3	3	3	1	-	-	-	-	-	-	-	3	3	2
3	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3
4	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3
5	3	3	3	3	1	-	-	-	-	-	-	-	3	3	3

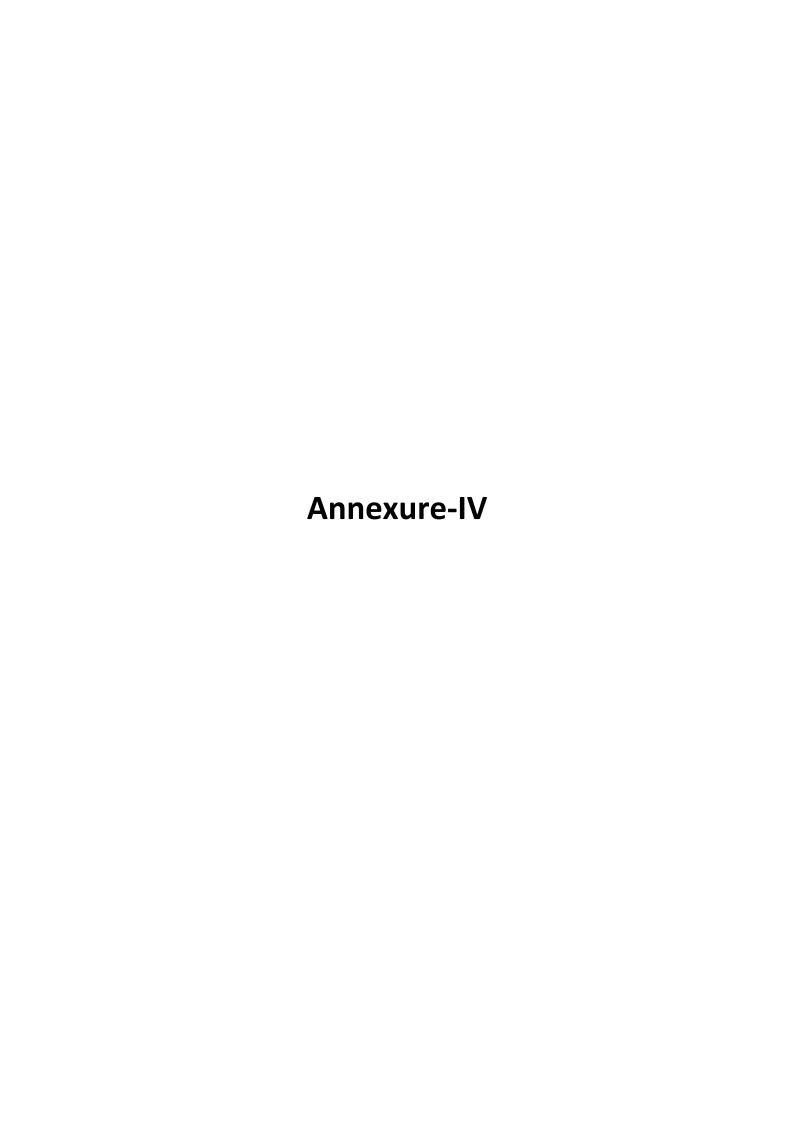
Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

	aiaatio		<u> </u>							
		Contin	uous Ass	essment Marks	(CAM)	<i>'</i>				
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Total Marks			
Marks	1	10 5 5		5	75	100				

Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

5-13/-



COLLEGE VISION AND MISSION

VISION

To be globally recognized for excellence in quality education, innovation and research for the transformation of lives to serve the society.

MISSION

M1: Quality Education : To provide comprehensive academic system that amalgamates

the cutting edge technologies with best practices.

M2: Research and Innovation : To foster value-based research and innovation in collaboration

with industries and institutions globally for creating

intellectuals with new avenues.

M3: Employability and

Entrepreneurship: To inculcate the employability and entrepreneurial skills

through value and skill based training.

M4: Ethical Values : To instill deep sense of human values by blending societal

righteousness with academic professionalism for the growth

of society.

DEPARTMENT VISION AND MISSION

VISION

Vision

Incorporating the Data Science skills and applying the acquired analytical knowledge in the heterogeneous domains through Artificial Intelligence

Mission

M1: Understand Data Science:

Amalgamation of Programming Knowledge, Mathematical Skill Set and Knowledge of Business Domains to face the challenges of the real-world requirement

M2: Applying the Acquired Knowledge:

Inculcating the spirit of applying the acquired knowledge, innovation and creativity among students to work in heterogeneous domains

M3: Capstone Project:

Providing forum to carry out a capstone project through collaborations with the industries

M4: Be socially beneficial and other moral concerns:

Inspiring the educational experience in the field of application development and ensure the design, principle and ethic to be followed in the society.

2.18/-

M5: Continuous Learning for keen Initiative:

Affording continuous learning in the field of current trends in Artificial Intelligence and Data Science for keen initiative and enterprise focused.

PROGRAMME OUTCOMES (POs)

PO1: Exploration of Research: An ability to independently carry out research/investigation and development work to solve practical problems.

PO2: Technical Skill: An ability to write and present a substantial technical report/document.

PO3: Expertise in Academics: Students should be able to demonstrate a degree of mastery over the area as per the specialization of the program. The mastery should be at a level higher than the requirements in the appropriate bachelor program.

PO4: Scholarship of Knowledge: Acquire in-depth knowledge of specific discipline or professional area, including wider and global perspective, with an ability to discriminate, evaluate, analyze and synthesize existing and new knowledge, and integration of the same for enhancement of knowledge.

PO5: Usage of Modern Tools: Create, select, learn and apply appropriate techniques, resources, and modern engineering and IT tools, including prediction and modelling, to complex engineering activities with an understanding of the limitations.

PO6: Ethical Practices and Social Responsibility: Acquire professional and intellectual integrity, professional code of conduct, ethics of research and scholarship, consideration of the impact of research outcomes on professional practices and an understanding of responsibility to contribute to the community for sustainable development of society.

PROGRAM EDUCATIONAL OBJECTIVES (PEOs)

PEO1: Competitive Platform: To create a competitive platform for solving critical problems in a wide variety of fields.

PEO2: Exploration: Enthusiastic participation in learning, understanding, designing and applying new innovative research ideas as the field evolves.

PEO3: Career: Applying cutting-edge technology that improves knowledge and to commit students for lifelong learning to reach the leading positions in the career.

PEO4: Professional Values: Simulate the graduates to hold the responsibilities in the context of technology, ethics, society and humanity.

2.18/-

PROGRAM SPECIFIC OUTCOMES (PSOs)

PSO1: Computational Skills: Graduates with the ability to apply basic knowledge of Computer Science in solving critical problems.

PSO2: Studious Research: Ability to convert innovative ideas into research or society-oriented projects through current trending technologies.

PSO3: Employability: Acquire placement in highly reputed industries or accomplish new technical business skills with the contemporary trends in the industry.

STRUCTURE FOR POSTGRADUATE ENGINEERING PROGRAM

SI.No	Course Category	Breakdown of Credits
1	Humanities and Social Sciences (HS)	6
2	Basic Sciences (BS)	3
3	Engineering Sciences (ES)	-
4	Professional Core (PC)	25
5	Professional Electives (PE)	18
6	Open Electives (OE)	-
7	Professional Activities (PA)	20
8	Ability Enhancement Courses (AEC)	-
9	Mandatory courses (MC)	-
	Total	72



SCHEME OF CREDIT DISTRIBUTION - SUMMARY

SUNG	Course Cotonomi		Credi Sem	r	Total	
SI.No	Course Category	ı	II	Ш	IV	Credits
1	Humanities and Social Sciences (HS)	4	2	-	-	6
2	Basic Sciences (BS)	3	-	-	-	3
3	Engineering Sciences (ES)	-	-	-	-	-
4	Professional Core (PC)	11	14	-	-	25
5	Professional Electives (PE)	3	6	9	-	18
6	Open Electives (OE)	-	-	-	-	-
7	Professional Activities (PA)			8	12	20
8	Ability Enhancement Courses (AEC)*	-	-	-	-	-
9	Mandatory Courses (MC)*	-	-	-	-	-
	Total	21	22	17	12	72

^{*} AEC, MC Credits are not included for CGPA calculation

5. NS/--

CURRICULUM

		SEI	MESTER - I							
SI.	Course Code	Course Title	Cotogony	Pe	erio	ds	Cradita	М	ax. Mar	ks
No.	Course Code	Course Title	Category	L	T	Р	Credits	CAM	ESM	Total
Thec	ry				,					
1	P23MAT105	Probability and Statistics	BS	2	1	0	3	40	60	100
2	P23ADT101	Machine Learning Algorithms	PC	3	0	0	3	40	60	100
3	P23ADT102	Computing Systems for Data Science	PC	3	0	0	3	40	60	100
4	P23ADT103	Artificial Intelligence and Intelligent Systems	PC	3	0	0	3	40	60	100
5	P23HSTC01	Research Methodology and IPR	HS	2	0	0	2	40	60	100
6	P23ADE1XX	Professional Elective – I *	PE	3	0	0	3	40	60	100
Prac	tical									
7	P23ADP101	Machine Learning Algorithms Laboratory	PC	0	0	4	2	50	50	100
8	P23HSPC01	Technical Report Writing and Seminar	HS	0	0	4	2	100	-	100
Abili	ty Enhancement	Course			•					•
9	P23ADC1XX	Ability Enhancement Course-I #	AEC	0	0	4	-	100	-	100
10	P23ACT10X	Audit Course-I**	AEC	0	0	2	-	100	-	100
	<u>-</u>	<u> </u>					21	590	410	1000

		SEMES	TER – II							
SI.	Course	Course Title	Category	Pe	erio	ds	Credits	Ma	ax. Mar	ks
No.	Code	Course Title	Category	L	T	Р	Credits	CAM	ESM	Total
The	ory									
1	P23ADT204	Parallel Programming Paradigms	PC	3	0	0	3	40	60	100
2	P23ADT205	Natural Language Processing	PC	3	0	0	3	40	60	100
3	P23ADT206	Advanced Deep Learning	PC	3	0	0	3	40	60	100
4	P23ADT207	Al and Robotic Process Automation	PC	3	0	0	3	40	60	100
5	P23ADE2XX	Professional Elective - II	PE	3	0	0	3	40	60	100
6	P23ADE2XX	Professional Elective - III	PE	3	0	0	3	40	60	100
Prac	tical									
7	P23ADP202	Advanced Deep Learning Laboratory	PC	0	0	4	2	50	50	100
8	P23HSPC02	Seminar on ICT: A Hands-on Approach	HS	0	0	4	2	100	ı	100
Abili	ity Enhanceme	nt Course								
9	P23ADC2XX	Ability Enhancement Course-II #	AEC	0	0	4	-	100	-	100
10	P23ACT20X	Audit Course-II**	AEC	0	0	2	-	100	-	100
							22	590	410	1000



		SEMES	STER – III							
SI.	Course	Course Title	Category	Р	erio	ds	Credits	Ma	ax. Mar	ks
No.	Code	Course Title	Category	L	T	Р	Credits	CAM	ESM	Total
Thec	ory									
1	P23ADE3XX	Professional Elective – IV *	PE	3	0	0	3	40	60	100
2	P23ADE3XX	Professional Elective – V *	PE	3	0	0	3	40	60	100
3	P23ADE3XX	Professional Elective – VI *	PE	3	0	0	3	40	60	100
Prac	tical									
7	P23ADW301	Project Phase - I	PA	0	0	12	6	50	50	100
8	P23ADW302	Internship	PA	0	0	0	2	100	-	100
Abili	Ability Enhancement Course									
10	P23ADC301	NPTEL/SWAYAM/MOOC	AEC	0	0	0	-	100	-	100
	·		·				17	370	230	600

	SEMESTER – IV									
SI.	Course	Course Title	Catagory	Periods			Credits	Max. Marks		
No.	Code	Course Title	Category	L	T	Ρ	Credits	CAM	ESM	Total
Prac	tical									
7	P23ADW403	Project Phase - II	PA	0	0	24	12	50	50	100
	12 50 50 100									

^{*} Professional Elective Courses are to be selected from the list given in Annexure I # Ability Enhancement Courses are to be selected from the list given in Annexure II

BS-Basic Sciences

PC - Professional Core

PE - Professional Elective

HS - Humanities and Social Sciences

PA - Professional Activity

CC- Common Course

AC- Audit Course

AEC - Ability Enhancement Course

CREDIT DISTRIBUTION

Semester	ı	II	III	IV	Total
Credits	21	22	17	12	72

Total number of credits required to complete

M.Tech in Computer Science and Engineering : 72 credits

2.18/-

^{**} Audit Courses are to be selected from the list given in Annexure III

ANNEXURE- I PROFESSIONAL ELECTIVE COURSES

SI. No. Course Code Course Title	PROFESSIONAL ELECTIVE COURSES						
1 P23ADEC01 Agile and Software Project Management 2 P23ADE101 Python for Data Science 3 P23ADE102 Data Science Essentials 4 P23ADE103 Big Data Mining and Analytics 5 P23ADE104 Artificial Intelligence for Decision Making Professional Elective-II 1 P23BDEC02 Web Analytics and Development 2 P23ADE205 Data Visualization using Tableau and Power BI 3 P23ADE206 Predictive Modelling 4 P23ADE207 Next Generation Database Systems 4 P23ADE208 Advanced Algorithms Professional Elective-III 1 P23ADE209 Al and Robotics 2 P23ADE210 Explainable Artificial Intelligence 3 P23ADE211 Introduction to Real-time Data Analytics 4 P23ADE212 Data Engineering in the Cloud 5 P23ADE213 Machine learning on Cloud platform Professional Elective-IV 1 P23ADE314 Generative Adversarial Networks 2 P23ADE315 Introduction to Large Language Models (LLMs) 3 P23ADE316 Transfer Learning 4 P23ADE317 Information Retrieval and Text Mining 5 P23ADE318 Statistical Natural Language Processing (NLP) Professional Elective-V 1 P23ADE319 Predictive Analytics 3 P23ADE320 Prescriptive Analytics 4 P23ADE321 Descriptive analytics 5 P23ADE322 Internet of Things (IoT) Data Analytics 5 P23ADE323 Social Media Analytics Professional Elective-VI 1 P23ADE324 Machine Learning Model Deployment and Management 2 P23ADE325 Al for Sustainability Professional Elective-VI 1 P23ADE324 Machine Learning Model Deployment and Management 2 P23ADE325 Al for Sustainability 3 P23ADE327 Al Applications in Cloud Computing	SI. No.	Course Code	Course Title				
2 P23ADE101 Python for Data Science 3 P23ADE102 Data Science Essentials 4 P23ADE103 Big Data Mining and Analytics 5 P23ADE104 Artificial Intelligence for Decision Making Professional Elective-II 1 P23BDEC02 Web Analytics and Development 2 P23ADE205 Data Visualization using Tableau and Power BI 3 P23ADE206 Predictive Modelling 4 P23ADE207 Next Generation Database Systems 5 P23ADE208 Advanced Algorithms Professional Elective-III 1 P23ADE209 Al and Robotics 2 P23ADE210 Explainable Artificial Intelligence 3 P23ADE211 Introduction to Real-time Data Analytics 4 P23ADE212 Data Engineering in the Cloud 5 P23ADE213 Machine learning on Cloud platform Professional Elective-IV 1 P23ADE214 Generative Adversarial Networks 2 P23ADE315 Introduction to Large Language Models (LLMs) 3 P23ADE316 Transfer Learning 4 P23ADE317 Information Retrieval and Text Mining 5 P23ADE319 Predictive Analytics 2 P23ADE319 Predictive Analytics 3 P23ADE320 Prescriptive Analytics 5 P23ADE321 Descriptive Analytics 5 P23ADE322 Internet of Things (IoT) Data Analytics 5 P23ADE323 Social Media Analytics 5 P23ADE324 Machine Learning Model Deployment and Management 1 P23ADE324 Machine Learning Model Deployment and Management 2 P23ADE325 Al for Sustainability 3 P23ADE326 Al in Natural Language Processing 4 P23ADE327 Al Applications in Cloud Computing	Profess	ional Elective-I					
3 P23ADE102 Data Science Essentials 4 P23ADE103 Big Data Mining and Analytics 5 P23ADE104 Artificial Intelligence for Decision Making Professional Elective-II 1 P23BDEC02 Web Analytics and Development 2 P23ADE205 Data Visualization using Tableau and Power BI 3 P23ADE206 Predictive Modelling 4 P23ADE207 Next Generation Database Systems 5 P23ADE208 Advanced Algorithms Professional Elective-III 1 P23ADE209 Al and Robotics 2 P23ADE210 Explainable Artificial Intelligence 3 P23ADE211 Introduction to Real-time Data Analytics 4 P23ADE212 Data Engineering in the Cloud 5 P23ADE213 Machine learning on Cloud platform Professional Elective-IV 1 P23ADE314 Generative Adversarial Networks 2 P23ADE315 Introduction to Large Language Models (LLMs) 3 P23ADE316 Transfer Learning 4 P23ADE317 Information Retrieval and Text Mining 5 P23ADE318 Statistical Natural Language Processing (NLP) Professional Elective-V 1 P23ADE319 Predictive Analytics 2 P23ADE320 Prescriptive Analytics 5 P23ADE321 Descriptive analytics 5 P23ADE321 Descriptive analytics 5 P23ADE323 Social Media Analytics 5 P23ADE324 Machine Learning Model Deployment and Management 2 P23ADE324 Machine Learning Model Deployment and Management 2 P23ADE327 Al Applications in Cloud Computing	1	P23ADEC01	1				
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5 P23ADE318 Statistical Natural Language Processing (NLP) Professional Elective-V 1 P23ADE319 Predictive Analytics 2 P23ADE320 Prescriptive Analytics 3 P23ADE321 Descriptive analytics 4 P23ADE322 Internet of Things (IoT) Data Analytics 5 P23ADE323 Social Media Analytics Professional Elective-VI 1 P23ADE324 Machine Learning Model Deployment and Management 2 P23ADE325 Al for Sustainability 3 P23ADE326 Al in Natural Language Processing 4 P23ADE327 Al Applications in Cloud Computing	3	P23ADE316	Transfer Learning				
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Professional Elective-VI 1 P23ADE324 Machine Learning Model Deployment and Management 2 P23ADE325 Al for Sustainability 3 P23ADE326 Al in Natural Language Processing 4 P23ADE327 Al Applications in Cloud Computing	4	P23ADE322	Internet of Things (IoT) Data Analytics				
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1 P23ADE324 Machine Learning Model Deployment and Management 2 P23ADE325 AI for Sustainability 3 P23ADE326 AI in Natural Language Processing 4 P23ADE327 AI Applications in Cloud Computing	Profession						
3 P23ADE326 Al in Natural Language Processing 4 P23ADE327 Al Applications in Cloud Computing			Machine Learning Model Deployment and Management				
4 P23ADE327 Al Applications in Cloud Computing	2	P23ADE325	Al for Sustainability				
P22ADE228 Ethics in All and Data Science	3	P23ADE326	Al in Natural Language Processing				
5 P23ADE328 Ethics in AI and Data Science	4	P23ADE327	Al Applications in Cloud Computing				
<u> </u>	5	P23ADE328	Ethics in AI and Data Science				



ANNEXURE- II ABILITY ENHANCEMENT COURSES

SI. No.	Course Code	Course Title	
1	P23XXCX01	Adobe Photoshop	
2	P23XXCX02	Adobe Animate	
3	P23XXCX03	Adobe Dreamweaver	
4	P23XXCX04	Adobe After Effects	
5	P23XXCX05	Adobe Illustrator	
6	P23XXCX06	Adobe InDesign	
7	P23XXCX07	Autodesk AutoCAD -ACU	
8	P23XXCX08	Autodesk Inventor - ACU	
9	P23XXCX09	Autodesk Revit - ACU	
10	P23XXCX10	Autodesk Fusion 360 - ACU	
11	P23XXCX11	Autodesk 3ds Max - ACU	
12	P23XXCX12	Autodesk Maya - ACU	
13	P23XXCX13	Cloud Security Foundations	
14	P23XXCX14	Cloud Computing Architecture	
15	P23XXCX15	Cloud Foundation	
16	P23XXCX16	Cloud Practitioner	
17	P23XXCX17	Cloud Solution Architect	
18	P23XXCX18	Data Engineering	
19	P23XXCX19	Machine Learning Foundation	
20	P23XXCX20	Robotic Process Automation / Medical Robotics	
21	P23XXCX21	Advance Programming Using C	
22	P23XXCX22	Advance Programming Using C ++	
23	P23XXCX23	C Programming	
24	P23XXCX24	C++ Programming	
25	P23XXCX25	CCNP Enterprise: Advanced Routing	
26	P23XXCX26	CCNP Enterprise: Core Networking	
27	P23XXCX27	Cisco Certified Network Associate - Level 2	
28	P23XXCX28	Cisco Certified Network Associate- Level 1	
29	P23XXCX29	Cisco Certified Network Associate- Level 3	
30	P23XXCX30	Fundamentals Of Internet of Things	
31	P23XXCX31	Internet Of Things / Solar and Smart Energy System with IoT	



	T	
32	P23XXCX32	Java Script Programming
33	P23XXCX33	NGD Linux Essentials
34	P23XXCX34	NGD Linux I
35	P23XXCX35	NGD Linux II
36	P23XXCX36	Advance Java Programming
37	P23XXCX37	Android Programming / Android Medical App Development
38	P23XXCX38	Angular JS
39	P23XXCX39	Catia
40	P23XXCX40	Communication Skills for Business
41	P23XXCX41	Coral Draw
42	P23XXCX42	Data Science Using R
43	P23XXCX43	Digital Marketing
44	P23XXCX44	Embedded System Using C
45	P23XXCX45	Embedded System with IOT / Arduino
46	P23XXCX46	English For IT
47	P23XXCX47	Plaxis
48	P23XXCX48	Sketch Up
49	P23XXCX49	Financial Planning, Banking and Investment Management
50	P23XXCX50	Foundation Of Stock Market Investing
51	P23XXCX51	Machine Learning / Machine Learning for Medical Diagnosis
52	P23XXCX52	IOT Using Python
53	P23XXCX53	Creo (Modelling & Simulation)
54	P23XXCX54	Soft Skills, Verbal, Aptitude
55	P23XXCX55	Software Testing
56	P23XXCX56	MX-Road
57	P23XXCX57	CLO 3D
58	P23XXCX58	Solid works
59	P23XXCX59	Staad Pro
60	P23XXCX60	Total Station
61	P23XXCX61	Hydraulic Automation
62	P23XXCX62	Industrial Automation
63	P23XXCX63	Pneumatics Automation
64	P23XXCX64	Agile Methodologies
65	P23XXCX65	Block Chain
66	P23XXCX66	Devops
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67	P23XXCX67	Artificial Intelligence
68	P23XXCX68	Cloud Computing
69	P23XXCX69	Computational Thinking
70	P23XXCX70	Cyber Security
71	P23XXCX71	Data Analytics
72	P23XXCX72	Databases
73	P23XXCX73	Java Programming
74	P23XXCX74	Networking
75	P23XXCX75	Python Programming
76	P23XXCX76	Web Application Development (HTML, CSS, JS)
77	P23XXCX77	Network Security
78	P23XXCX78	MATLAB
79	P23XXCX79	Azure Fundamentals
80	P23XXCX80	Azure AI (AI-900)
81	P23XXCX81	Azure Data (DP -900)
82	P23XXCX82	Microsoft 365 Fundamentals (SS-900)
83	P23XXCX83	Microsoft Security, Compliance and Identity (SC-900)
84	P23XXCX84	Microsoft Power Platform (Pl-900)
85	P23XXCX85	Microsoft Dynamics Fundamentals 365 – CRM
86	P23XXCX86	Microsoft Excel
87	P23XXCX87	Microsoft Excel Expert
88	P23XXCX88	Securities Market Foundation
89	P23XXCX89	Derivatives Equinity
90	P23XXCX90	Research Analyst
91	P23XXCX91	Portfolio Management Services
92	P23XXCX92	Cyber Security
93	P23XXCX93	Cloud Security
94	P23XXCX94	PMI – Ready
95	P23XXCX95	Tally – GST & TDS
96	P23XXCX96	Advance Tally
97	P23XXCX97	Associate Artist
98	P23XXCX98	Certified Unity Programming
99	P23XXCX99	VR Development



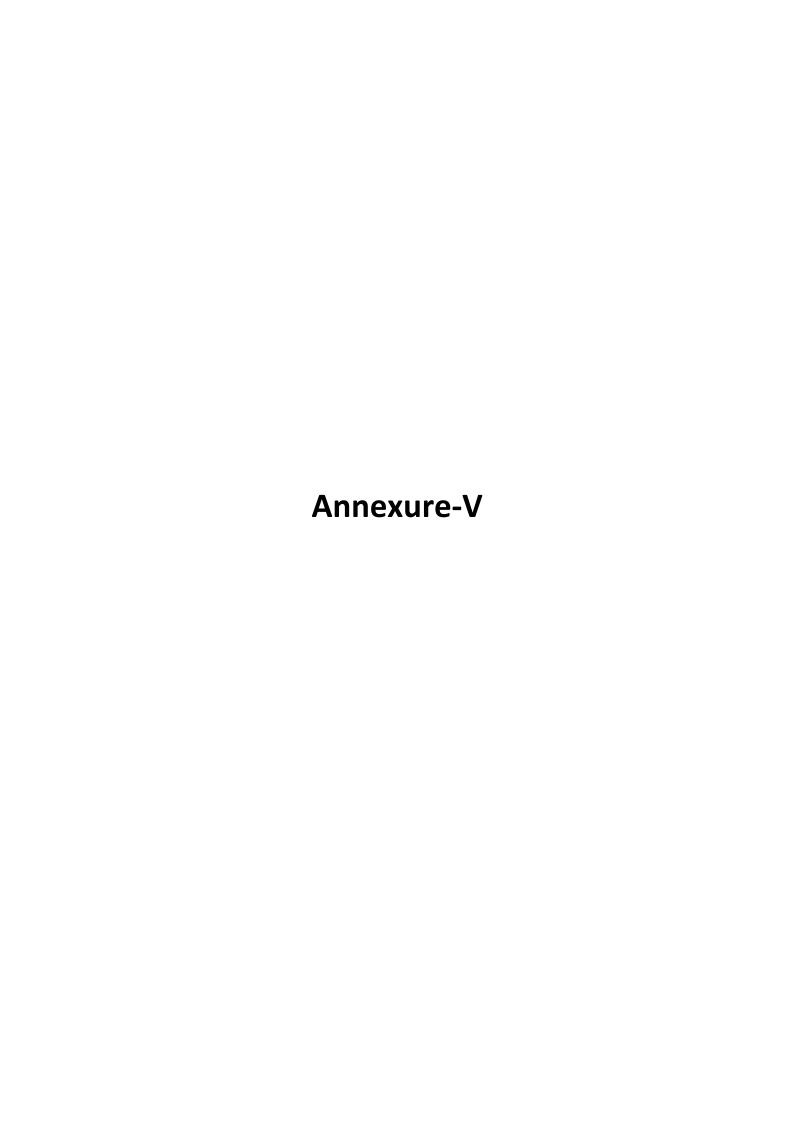
ANNEXURE-III

AUDIT COURSES

(Common to all M.Tech Programme)

SI. No.	Course Code	Course Title
1	P23ACTX01	English for Research Paper Writing
2	P23ACTX02	Disaster Management
3	P23ACTX03	Sanskrit for Technical Knowledge
4	P23ACTX04	Value Education
5	P23ACTX05	Constitution of India
6	P23ACTX06	Pedagogy Studies
7	P23ACTX07	Stress Management by Yoga
8	P23ACTX08	Personality Development Through Life Enlightenment Skills
9	P23ACTX09	Unnat Bharat Abhiyan

5. X S/ ._



Department	Mathe	natics		Progran	nme : N	I.Tech.				
Semester	I			Course	Catego	ry: BS	*Er	nd Semest	erExamTy _l	pe: TE
Course Code	P23M <i>A</i>	T105		Perio	ds/We	ek .	Credit	Max	kimumMarl	ks
				L	Т	Р	С	CAM	ESE	TM
Course Name	PROB	ABILITY	AND STATISTICS	2	1	0	3	40	60	100
		(AI&I	/							
Prerequisite		Mathem							·····	
	On co	ompletic	on of the course, the stu	dents will b	e able	to			BT Ma (Highest	
	CO1	Apply the	e concept of probability in rar	ndom variable	S.				(Filightes)	······································
		Apply the	e basic rules of continuous ra	andom variabl	es.				K	3
Course	CO3	Apply the	e concept of testing of hypoth	hesis for small	and lar	ge samp	les in real life	problems.	K	2
Outcome			oncept of linear regression, c			•			K	
	CO5	List the c	guidelines for designing expe	eriments and re	ecoanize	e the key	historical figu	ures in	K	
			f Experiments.		3	- · · · · · ·	3			
UNIT – I	DISC	RETE R	ANDOM VARIABLES				Periods:12	2	·····	
		eir event s	paces – The probability mas	s function – Di	stributio	n functio	ns – Binomial	– Geometri	c – Negative	
Binomial and Poi	sson.									CO1
UNIT – II	CONT	INUOU	S RANDOM VARIABLES	3			Periods:12	2		i.
Some important Reliability – Failu			oonential distribution –Gamr zard function.	ma – Weibull	– Gauss	sian disti	ibutions. App	lication of o	distribution -	CO2
UNIT – III	TEST	ING OF	HYPOTHESIS				Periods:12	2		<u>i</u>
Sampling distribu	tions – S	mall and	large samples –Tests based	on Normal, t t	est, Chi	square t	est, and F tes	t distributio	ns for testing	g
of means, variand	ce and pi	oportions	s — Contingency table (test	for independe	nt) Good	dness of	fit.			CO3
UNIT – IV	<u>. i</u>		ON AND REGRESSION				Periods:12			
			gression –Multiple and part Coefficient of partial correlat		Meth	od of le	ast squares -	- Plane of	regression -	- CO4
UNIT – V	DESI	GN OF E	EXPERIMENTS				Periods:12	2		
Analysis of variar square design - 2			d two-way classifications – 0	Completely ra	ndomize	ed desigr	n – Randomiz	ed block de	esign – Latii	CO5
LecturePerio	ds:45		TutorialPeriods:15	Practica	alPerio	ds: -	Т	otalPerio	ds:60	<u>i</u>
Text Books				•			-			
			ng Mathematics", Khanna ρι	-						
•			tistics and Random Processe							
•		oor, V.K	., "Fundamentals of Mathem	atical Statistic	s", Sulta	ın Chanc	l and Sons, 1	2th Edition,	2023.	
Reference Boo		Db -# "F		-C	Г.J:и:	2047				
			ngineering Mathematics", Mo				otiotico" Con			
2. William Mende 15 th Edition 201		beil J. D	eaver, Barbara M. Beaver: "I	miroduction to	Probab	iiity & St	ausucs , cenç	gage Learn	rig,	
			and John E. Freund," Proba	ability and Stat	istics fo	r Engine	ers",			
Pearson Educa 4. Vijav K. Rohat			tion, 2018. nsanesSaleh, "An Introductio	on to Probabili	tv and S	tatistics"	. Wilev – 200	8.		
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5. <u>E. Rukm</u> angad	-		y and Statistics", Pearson E	ducation India	, 2012.					
5. <u>E. Rukmangac</u> Web Referenc e	lachari, "			ducation India	, 2012.					
	lachari, " es	Probabilit		ducation India	, 2012.					
Web Reference 1. http://www.s	lachari, " es stat110.n	Probabilit et		ducation India	, 2012.					
Web Reference 1. http://www.s	lachari, " es stat110.n ptel.ac.in	Probabilit et /courses/	y and Statistics", Pearson Ed	ducation India	, 2012.					
Web Reference 1. http://www.s 2. http://www.n	lachari, " es stat110.n ptel.ac.in probabilit	Probabilit et /courses/ /course.c	y and Statistics", Pearson Ed	ducation India	, 2012.					
Web Reference 1. http://www.s 2. http://www.n 3. http://www.p 4. www.edx.org	lachari, " es stat110.n ptel.ac.in probabilit	Probabilit et /courses/ /course.c	y and Statistics", Pearson Ed	ducation India	, 2012.					

^{*} TE – Theory Exam, LE – Lab Exam

COs		Progra	m Out	Program Specific Outcomes (PSOs)					
	PO1	PO2	PO3	PO4	PO5	P06	PSO1	PSO2	PSO3
1	2	1	-	-	-	1	1	2	1
2	3	2	1	1	-	1	2	2	1
3	3	2	1	1	-	1	2	2	1
4	3	2	1	1	-	ı	2	2	1
5	3	2	1	1	-	-	2	2	1

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	(CAM)	End Semester	Total	
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

^{**}Assignment to be given from Unit-5

	Artific	ial Intelligence and Data Science	Progran	nme : N	I.Tech.					
Semester	I		Course Category: PC *End Seme						terExamT	ype: TE
Course Code			Perio	ds/We	ek	Cre	dit	Ma	ximumMa	rks
Course Code	P23AD	DT101	L	Т	Р	С	•	CAM	ESE	TM
Course Name	MACH	INE LEARNING ALGORITHMS	3	0	0	3	3	40	60	100
		(AI&DS)								
Prerequisite	NIL				-					
	On co	ompletion of the course, the stude	nts will b	e able	to					apping
Course Outcome	CO1	Interpret the concepts of machine learni	ng algorith	ms and	use kno	wledge o	n regr	ession		st Level (2
Odtcome	CO2	Apply the classification and the clustering	ng algorithn	ns for the	eir mod	els			ŀ	(3
	CO3	Apply the dimensionality reduction tech	nigues in th	eir mod	els				ŀ	(3
	<u> </u>	Apply Decision learning algorithm and re	-			models			ŀ	(3
	CO5	Identify the use of ensemble learning armodels	nd apply the	em to im	prove th	ne efficien	cy of	their		, K 3
UNIT – I	Intro	duction to Machine Learning				Period	ds:9			
	e Learni	ng - Types of Machine Learning Algorith		Preproce	essing -	Introducti	on of	Regression	n Algorithn	ns
- Linear Regress	ion – Mu	ıltivariate Linear Regression – Logistic R	egression.							CO1
UNIT – II	Class	sification and Clustering Models				Period	ls:9			
		gorithms – Support Vector Machine – Na	aive Bayes	– classif	ying wit	<u>.i.</u>		robabilities	s – K-Neare	est CO2
		stering Algorithms - K-Means clustering -								
UNIT – III	Dime	nsionality Reduction Techniques				Period	ls:9			
_		-				1				
		ection - Principal Component Analysis	(PCA) -	Factor a	analysis	Multid	imens	sional Sca	iling - Line	ar
			(PCA) -	Factor a	analysis	Multid	imens	sional Sca	lling - Line	
Discriminant Ana	lysis (LD	A) Case Study.		Factor a	analysis	·•		sional Sca	lling - Line	
Discriminant Ana	lysis (LD		ng			Period	ds:9			CO3
Discriminant Ana UNIT - IV Decision tree rep	Decis Decisoresentat	A) Case Study. sion Trees and Rule Based Learnir	ng			Period	ds:9			CO3
Discriminant Ana UNIT – IV Decision tree reposith Apriori and E	Decisoresentat Equivaler	A) Case Study. Sion Trees and Rule Based Learnir ion – ID3 – CART – Hidden Morkov Monce Class Transformation Algorithm	ng			Perioc	ds:9 ociatio			CO3
Discriminant Ana UNIT – IV Decision tree reposith Apriori and E UNIT – V	Decis presentat Equivaler	A) Case Study. Sion Trees and Rule Based Learnir ion – ID3 – CART – Hidden Morkov Mo	ng del - Asso	ciation ru	ule mini	Period	ds:9 ociatio	on rules -		CO3
Discriminant Ana UNIT – IV Decision tree reposith Apriori and E UNIT – V Introduction – Ba	Decis presentat Equivaler Ense gging: R	A) Case Study. Sion Trees and Rule Based Learnir ion – ID3 – CART – Hidden Morkov Monce Class Transformation Algorithm mble Learning andom Forest – Boosting: Adaboost and	ng del - Associ	ciation ru	ule mini	Period	ds:9 ociatio ds:9 Stackii	on rules -	Case studi	CO3
Discriminant Ana UNIT – IV Decision tree reposith Apriori and E UNIT – V Introduction – Ba LecturePerior	Decis presentat Equivaler Ense gging: R	A) Case Study. Sion Trees and Rule Based Learnir ion – ID3 – CART – Hidden Morkov Mo nce Class Transformation Algorithm mble Learning	ng del - Asso	ciation ru	ule mini	Period	ds:9 ociatio ds:9 Stackii	on rules -	Case studi	CO3
UNIT – IV Decision tree reports April 1 – IV UNIT – V UNIT – V UNIT – V UNIT – Ba LecturePerion	Decis presentat Equivaler Ense gging: R ds:45	ion Trees and Rule Based Learning ion – ID3 – CART – Hidden Morkov Monce Class Transformation Algorithm imble Learning andom Forest – Boosting: Adaboost and TutorialPeriods:0	ng del - Associ I XGBoost A	ciation ru	ule mini ns Light	Perioc ng – Asso Perioc GBM – S	ds:9 ociatio ds:9 Stackii	on rules - · ng. otalPerio	Case studio	CO3
Discriminant Ana UNIT – IV Decision tree reported from the Policy UNIT – V Introduction – Bath LecturePerion 1. Henrik E	Decisoresentatequivaler Ense gging: R ds:45	Sion Trees and Rule Based Learnir ion – ID3 – CART – Hidden Morkov Monce Class Transformation Algorithm mble Learning andom Forest – Boosting: Adaboost and TutorialPeriods:0	ng I XGBoost A Practic	ciation ru	ule mini ns Light	Perioc ng – Asso Perioc GBM – S	ds:9 ociatio ds:9 Stackii	on rules - · ng. otalPerio	Case studio	CO3
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https://www.coursera.org/learn/machine-learning.
 https://www.youtube.com/watch?v=Gwlo3gDZCVQ

COs	I	Progra	m Out	Program Specific Outcomes (PSOs)					
	P01	PO2	PO3	PO4	PO5	P06	PSO1	PSO2	PSO3
1	3	3	2	3	2	-	3	2	-
2	3	1	2	2	3	-	3	1	-
3	2	1	2	3	3	-	3	1	-
4	2	1	2	2	3	-	3	1	-
5	3	1	2	2	3	-	3	1	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	End Semester	Total		
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus **Assignment to be given from Unit-5

5. X } \ _

	Artific	ial Intelligence and Data Science	Program	me: M.	Tech.				
Semester	I		Course (Categoi	ry: PC	*End	Semester	Exam Ty	ре: ТЕ
Course Code	P23AI	T102	Period	ds / We	ek	Credit	Maxiı	num Mai	'ks
Course Code	1 2071	31102	L	Т	Р	С	CAM	ESE	TM
Course Name	COMP	PUTING SYSTEMS FOR DATA ICE	3	0	0	3	40	60	100
		(/	AI & DS)						
Prerequisite	Compu	ter Organization and Architecture/Op	erating Sy	/stem/D	Database	e Managem	ent System	S	
	On co	mpletion of the course, the studer	nts will be	able t	0				lapping st Level
Course Outcomes	CO1 Interpret the issues related to the design and analysis of systems with real-time constraints and to identify the features of Real time OS								
	CO2	Classify and Compare various Uniproc	essor and I	Multipro	cessor s	cheduling me	chanisms	l	K3
	CO3	Categorize the difference between trace	ditional and	real tim	e databa	ses.		l	K2
	CO4	Identify Data Storage and Managemer	nt Technolo	gies				K2	
	CO5	Interpret information about Storage Are	ea Network	s charac	cteristics	and compon	ents.	K2	
UNIT-I	Introd	uction				Periods: 9			
	de analys	ks - Hard and Soft timing constraints - De is, Microarchitecture level analysis, Cachime OS							
_	Threads a	and Tasks – Structure of Microkernel –	T::		<u>.</u>				
Synchronization -		otification and Software interrupt Task as or task scheduling - Clock-driven and pri	ssignment a	and Sch	eduling -	Task allocat	ion algorithm	s - Single	
Synchronization – processor and Mu	Itiprocess	otification and Software interrupt Task as	ssignment a	and Sch	eduling -	Task allocat	ion algorithm tolerant sche	s - Single	
Synchronization – processor and Mu UNIT-III Real time Databas	Real to ses - Tra	otification and Software interrupt Task as or task scheduling - Clock-driven and pri	ssignment a iority-based	and Sch I schedu	eduling - uling algo	Task allocat orithms Fault Periods: 9	ion algorithm tolerant sche	s - Single duling.	0
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Synchronization — processor and Mu UNIT-III Real time Database mprove predictab UNIT-IV Hard Disks- Netwo	Real t ses – Tra ility Large prized Atta	otification and Software interrupt Task as for task scheduling - Clock-driven and prisme Databases Insaction priorities - Concurrency control Data Storage Inched Storage-Scalability issues- Network	ssignment a iority-based I issues – E	and Sch I schedu Disk sch	eduling - uling algo eduling a	Task allocat rithms Fault Periods: 9 algorithms – Periods: 9	ion algorithm tolerant sche Fwo phase a	s - Single duling. pproach to	CO
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Synchronization — Processor and Mu UNIT-III Real time Database INIT-IV Hard Disks- Netwo System Design- C UNIT-V Storage Area Ne	Real to sees – Traility Large price orked Attaching-Le Storage tworks –	otification and Software interrupt Task as for task scheduling - Clock-driven and prisme Databases Insaction priorities – Concurrency control Data Storage Inched Storage-Scalability issues- Networkegacy Systems.	ssignment a iority-based I issues – C king issues	and Schedudischedu	eduling algo eduling a eduling a e Archite	Task allocated rithms Fault Periods: 9 Regional Periods: 9 cture - Storage Periods: 9 rage QoS- P	ion algorithm tolerant sche Two phase a ge Partitionin	s - Single duling. pproach to g- Storage Reliability	cc e cc

Lecture Periods: 30 Tutorial Periods: 15 Practical Periods: - Total Periods: 45

Text Books

- 1. C.M. Krishna, Kang G. Shin Real Time Systems, McGraw Hill Education; 1st Edition, 2017.
- 2. Jane W.S. Liu, Real-Time Systems", Pearson Education India, 2002.
- 3. Sanjoy Baruah, Marko Bertogna, Giorgio Buttazzo, Multiprocessor Scheduling for Real-Time Systems, Springer International Publishing, 2015.

Reference Books

- 1. Hermann Kopetz, Real-Time Systems: Design Principles for Distributed Embedded Applications, 4th Edition, 2011.
- 2. Robert Spalding and Daniel J. Worden, Storage Networks: The Complete Reference, 1st edition, 2003.
- 3. Phillip A. Laplante, Real-Time Systems Design and Analysis: Tools for the Practitioner, 3rd edition, 2004.
- 4. Qing Li, Caroline Yao, Real-Time Concepts for Embedded Systems, 1st edition, 2003.
- 5. Christopher Poelker, Alex Nikitin, Storage Area Networks for Dummies, 2nd edition, 2009.

Web References

- https://datascience.columbia.edu/research/centers/computing-systems-for-data-driven-science/https://www.guru99.com/what-is-tableau.html
- 2. https://www.geeksforgeeks.org/real-time-operating-system-rtos/
- 3. https://hazelcast.com/glossary/real-time-database/
- 4. https://www.simplilearn.com/big-data-storage-article
- 5. https://www.techtarget.com/searchstorage/definition/big-data-storage

* TE - Theory Exam, LE - Lab Exam

COs	I	Progra	m Out	comes	(POs)	Program Specific Outcomes (PSOs)			
	PO1	PO2	PO3	PO4	PO5	PO6	PSO1	PSO2	PSO3	
1	2	3	2	-	-	-	1	2	1	
2	3	3	3	-	1	-	2	2	1	
3	2	3	2	1	2	•	2	2	1	
4	3	3	3	1	3	1	2	2	1	
5	3	3	3	1	3	-	2	2	1	

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	End Semester	Total		
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus **Assignment to be given from Unit-5

	Aitiiit	ial Intelligence and Data Science	Program						
Semester	I		Course			*En	d Semester		
Course Code	P23AI	OT103	Perio	ds / We	ek	Credit		mum Ma	
			L	Т	Р	С	CAM	ESE	TM
Course Name		FICIAL INTELLIGENCE AND	3	0	0	3	40	60	100
	INIEL	LIGENT SYSTEMS	VI 0 DC/		<u> </u>				
Prerequisite	_	(4	AI & DS)						
rielequisite	On co	mpletion of the course, the stude	nts will be	e able t	0				1apping
Course	CO1	Apply different search techniques to so	olve real wo	orld prob	olems			÷	est Leve K3
Outcomes	CO2			•					
		Interpret different approaches of know						ļ	K2
	CO3	Apply the planning and learning appro					-		K3
	CO4	Interpret intelligent computing models can be solved					orld problems		K3
	CO5	Apply the benefits of hybridization and	propose ne	w hybrid	d algorith				K3
UNIT-I	<u>i</u>	uction and Search Techniques ng through search, state-space, blind se				Periods: 9			
	Know	ladge Penrocentation Techniques	and Pag	conina	undor	Pariade: 0	\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \		
pproaches for k esolution, Sema	uncer nowledge antic net	representation, Propositional Logic, Propositional	redicate Lo	gic, Rul	e based	knowledge	representation		
pproaches for k esolution, Sema robability Theory	uncer nowledge antic net , Bayes F	tainty representation, Propositional Logic, Proposi	redicate Lo	gic, Rul	e based	knowledge Resolution,	representation Managing U		
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- Crina Grosan and Ajith Abraham, Intelligent Systems- A Modern Approach, Springer Intelligent Systems Reference Library Book 17, 2011.
- 4. Michael Negnevitsky, Artificial Intelligence: A Guide to Intelligent Systems, 3rd edition, 2011.
- 5. Christopher M. Bishop, Pattern Recognition and Machine Learning, 1st Edition, 2006.

Web References

- 1. https://www.ibm.com/topics/data-science
- 2. https://www.oracle.com/in/what-is-data-science/
- 3. https://u-next.com/blogs/data-science/importance-of-data-science/
- 4. https://monkeylearn.com/text-analysis/
- $5. \quad \text{https://www.toptal.com/designers/data-visualization/data-visualization-tools} \\$

* TE – Theory Exam, LE – Lab Exam

2. 1 / / -

COs	İ	Progra	m Out	comes	s (POs) Program Specif Outcomes (PSO				
	PO1	PO2	PO3	PO4	PO5	P06	PSO1	PSO2	PSO3
1	3	3	2	3	2	-	3	2	-
2	3	3 1 2 2 3 -						1	-
3	2	1	2	3	3	-	3	1	-
4	2	1	2	2	3	-	3	1	-
5	3	1	2	2	3	-	3	1	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus
**Assignment to be given from Unit-5

Department	Artific	cial Intelligence and Data Science	nce Programme: M.Tech.						
Semester	I		Course	Catego	ry Code	e: HS *Er TE		ter Exam T	уре:
Course Code	DOSH	STC01	Perio	ods / We	eek	Credit	Ma	aximum Ma	rks
Course Code	газп	31001	L	Т	Р	С	CAM	ES E	ТМ
Course Name	RESE	ARCH METHODOLOGY AND IPR	2	0	0	2	40	60	100
Prerequisite	No pre	erequisite needed							
	On co	empletion of the course, the stude	nts will b	e able	to			BT Ma _l (Highest	
_	CO1	Interpret and formulate the research pro	oblem.					K2	2
Course Outcomes	CO2	Identify the concepts to carry out the lite	erature rev	/iew, eth	ics and	research anal	ysis.	K2	2
CO3 Identify the way of writing technical paper and presentation me						ethods. K2			
	CO4 Interpret the use of intellectual property rights							K2	
	CO5	File patents through Research and Dev	elopment/	cell				K3	3
UNIT-I	Resea	arch Problem Formulation				Periods: 6			
a research probler	n - scop	em- Sources of research problem - criteria be and objectives of research problem. A – interpretation - necessary instrumentati	Approache						CO1
UNIT-II	Litera	ture Review				Periods: 6			
Effective literature	studies	approaches - analysis - plagiarism and	research e	ethics					CO2
UNIT-III	Techr	nical Writing /Presentation				Periods: 6			
		 how to write report – paper - developing nent by a review committee. 	g a researd	ch propo	sal - forr	mat of researd	h proposa	l -	СОЗ
UNIT-IV	Introd	luction To Intellectual Property Rig	ghts (IPR	.)		Periods: 6			
		erty: patents – designs - trade and copyri							CO4
		atenting - development. International scel tents - patenting under pct.	nario: inter	national	coopera	ation on intelle	ctual prop	erty -	
UNIT-V		ectual Property Rights (IPR)				Periods: 6			<u> </u>
Patent Rights: So Indications - New	cope of l	Patent Rights - Licensing and transfer of opments in IPR - Administration of Patent ase Studies - IPR and IITs.				ation and data	abases - G		CO5
Lecture Period	······································	Tutorial Periods: -	Practic	al Perio	ods: -	Т	otal Peri	ods: 30	
Text Books						<u>L</u>			

- 1. Stuart Melville and Wayne Goddard, "Research methodology: An introduction for science & Engineering students', Kenwyn Publisher,
- 2. Wayne Goddard and Stuart Melville, "Research Methodology: An Introduction", Lansdowne Publisher, Second Edition, 2001.
- 3. C.R. Kothari, Gaurav Garg, "Research Methodology: Methods and Techniques", New Age International, Fourth Edition, 2018.

Reference Books

- 1. Halbert, "Resisting Intellectual Property", Taylor & Francis Limited, 2007.
- 2. Ranjit Kumar, "Research Methodology: A Step by Step Guide for beginners", Second Edition, 2010.
- 3. Trochim, "Research Methods: The concise knowledge base", Atomic Dog Publishing, 2005.
- 4. Fink A, "Conducting Research Literature Reviews: From the Internet to Paper", Sage Publications, 2009.

Web References

- 1. https://www.scribd.com/document/427419672/Research-Methodology-and-lpr
- 2. https://www.isical.ac.in/~palash/research-methodology/RM-lec9.pdf
- 3. https://www.wipo.int/edocs/pubdocs/en/intproperty/958/wipo_pub_958_3.pdf
- 4. https://lecturenotes.in/m/21513-research-methodology
- 5. https://iare.ac.in/sites/default/files/MTECH-CAD.CAM-R18-RM-IP-NOTES.pd

* TE - Theory Exam, LE - Lab Exam

COs	I	Progra	m Out	Program Specific Outcomes (PSOs)					
	PO1	PO2	PO3	P06	PSO1	PSO2	PSO3		
1	3	2	1	1	3	3	2		
2	3 2 1 1 2 1						3	2	2
3	3 2 1 1 2 1						3	2	2
4	3	2	1	1	3	2	3		
5	3	2	1	1	2	1	3	2	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High
* TE - Theory Exam, LE - Lab Exam

Evaluation Method

	(Contin	uous As	sessment Mark	s (CAM)	End	
Assessment	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Semester Examination (ESE) Marks	Total Marks
Marks	1	0	15	10	5	60	100

^{**}Assignment to be given from Unit-5

5. X } \ __

Department	Artificial Intelligence and Data Science	Progran	nme: M	.Tech				
Semester	I	Course	Catego	ry: PC		ind Semest .E	ter Exam	Туре:
Course Code	P23ADP101	Perio	ds / We	eek	Credit	Ma	ximum Ma	arks
Course Code	FZSADF 101	L	Т	Р	С	CAM	ESE	TM
Course Name	MACHINE LEARNING ALGORITHMS LABORATORY	0	0	4	2	50	50	100
	(<i>F</i>	Al & DS)	.4			<u>i</u>		
Prerequisite	Knowledge about Machine Learning Alg	orithms						
	On completion of the course, the stu-	dents will	be abl	e to				lapping st Level)
Course	CO1 Build models by applying Supervised	algorithms	•				l	K6
Outcomes	CO2 Build models by applying Unsupervis	ed algorithr	ns				I	K6
	CO3 Build models by applying Regression	Technique	s.				I	K6
	CO4 Build models by applying Dimensiona	lity Reducti	on techr	niques.			I	K6
	CO5 Build models by applying hybrid ense	emble algor	ithms				I	K6
List of Exercise	<u> </u>		•••••					

List of Exercises

- Support Vector Machine
- 2. Naive Bayes
- 3. K-Nearest Neighbor
- 4. Linear Regression and Logistic Regression
- 5. K-Means and K-Medians
- 6. Principal Component Analysis and Linear Discriminant Analysis
- 7. Decision Tree Algorithm
- 8. Naïve Bayes ensemble
- Random forests
- 10. Adaboost and XGBoost

				ı
Lecture Periods:	Tutorial Periods: -	Practical Periods: 45	Total Periods: 45	
n / n i				

Reference Books

- 1. Andreas C. Mueller and Sarah Guido, "Introduction to Machine Learning with Python", O'Reilly Media, Inc. First edition, 2016.
- 2. Henrik Brink, Joseph W. Richards, and Mark Fetherolf, "Real-World Machine Learning", Manning Publications, 2017.
- 3. Mehryar Mohri, Afshin Rostamizadeh, Ameet Talwalkar, "Foundations of Machine Learning", The MIT Press, 2nd Edition, 2009.

COs/POs/PSOs Mapping

COs	I	Progra	m Out	Program Specific Outcomes (PSOs)					
	PO1	PO2	PO3	P06	PSO1	PSO2	PSO3		
1	3	3	2	3	2	-			
2	3	1	2	2	3	-	3	1	-
3	2	1	2	3	3	-	3	1	-
4	2	1	2	1	3	1	-		
5	3	1	2	2	3	ı	3	1	2

Correlation Level: 1 - Low, 2 - Medium, 3 - High

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^{*} TE - Theory Exam, LE - Lab Exam

Evaluation Method

	Co	ntinuous <i>A</i>	ssess	ment Marks (CA	AM)		
Assessment	Performan cla	ce in pract asses	ical	Model		End Semester	Total
	Conduction of practical	Record work	viva	Practical Examination	Attendance	Examination (ESE) Marks	Marks
Marks	15	5	5	15	10	50	100

Department	Artificial intelligence and Data Science	Prograr	nme: M .	Tech.				
Semester	First	Course Category Code: HS *End Semester Exam					ter Exam	Гуре:
Course Code	P23HSPC01	Perio	ods / We	ek	Cred	it Ma	ximum Ma	arks
Course Code	F23113F601	L	Т	Р	С	CAM	ESE	TM
Course Name	TECHNICAL REPORT WRITING AND SEMINAR	0	0	4	2	100	-	100
(Co	mmon to all M.Tech Programme)							
Prerequisite				-				

Prerequisite			
	On co	ompletion of the course, the students will be able to	BT Mapping (Highest Level)
Course	CO1	Select a subject, narrowing the subject into a topic.	K2
Outcomes	CO2	State an objective and collecting the relevant bibliography (at least 15 journal papers).	K2
	CO3	Study the papers and understanding the author's contributions and critically analyzing each paper.	K3
	CO4	Prepare a working outline and linking the papers and preparing a draft of the paper.	K2
	CO5	Prepare a working outline and linking the papers and preparing a draft of the paper.	K2

List of Experiments:

Activity	Instructions	Submission week	Evaluation
Selection of area of interest and Topic	Select an area of interest, topic and state an objective	2 nd week	3 % Based on clarity of thought, current relevance and clarity in writing
Stating an Objective			
Collecting Information about area & topic	 List 1 Special Interest Groups or professional society List 2 journals List 2 conferences, symposia or workshops List 1 thesis title List 3 web presences (mailing lists, forums, news sites) List 3 authors who publish regularly in your area Attach a call for papers (CFP) from your area. 	3 rd week	3% (the selected information must be area specific and of international and national standard)
Collection of Journal papers in the topic in the context of the objective — collect 20 & then filte	 provide a complete list of references you will be using-Based on your objective -Search various digital libraries and Google Scholar When picking papers to read - try to: - Pick papers that are related to each other in some ways and/or that are in the same field so that you can write a meaningful survey out of them. Favour papers from well-known journals and conferences, in the field (as indicated in other Favour more recent papers, Pick a recent survey of the field so you can quickly gain an overview, Find relationships with respect to each other and to your topic area(classification scheme/categorization) Mark in the hard copy of papers whether complete work or section/sections of the paper are being considered 	4 th week	6% (the list of standard papers and reason for selection)
Reading and notes for first 5 papers	Reading Paper Process For each paper form a Table answering the following questions: • What is the main topic of the article? • What was/were the main issue(s) the author said they want to discuss? • Why did the author claim it was important? • What simplifying assumptions does the author claim to be making? • What did the author do? • How did the author claim they were going to evaluate their work and compare it to others? • What did the author say were the limitations of their research? • What did the author say were the important directions for future research?	6 th week	8% (The table given should indicate your understanding of the paper and the evaluation is based on your conclusions about each paper)

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Locturo Poriode:	Tutorial Pariade	Practical Pariods: 45	Total Pariods: 45
Seminar	A brief 15 slides on your paper	14th & 15th week	10% (based on presentation and Vivavoce)
Final Draft	Complete the final draft of your paper	13th week	10% (formatting, English, Clarity and linking) 4% Plagiarism Check Repor
Conclusions	Write your conclusions and future work	12th week	5% (conclusions)
Sections of the paper	Write the sections of your paper base classification / categorization diagram in keep the goals of your survey	eeping with	10% (this component will be evaluated based on the linking and classification among the papers)
Introduction Background	Write an introduction and background section		5% (clarity)
Abstract	Prepare a draft abstract and give a presenta		6%(Clarity, purpose and conclusion) 6% Presentation & Viva Voce
Draft outline 1 and Linking papers	Prepare a draft Outline, your survey goals, a classification / categorization diagram	along with a 8 th week	8% (this component will be evaluated based on the linking and classification among the papers)
Reading and notes for next 5 papers	Repeat Reading Paper Process	7 th week	8% (the table given should indicate your understanding of the paper and the evaluation is based on your conclusions about each paper
	 Conclude with limitations/issues not addre paper (from the perspective of survey) 	ssed by the	

Lecture Periods: - Tutorial Periods: - Practical Periods: 45 Total Periods: 45

COs/POs/PSOs Mapping

COs	I	Progra	m Out	Program Specific Outcomes (PSOs)							
	P01	PO2	PO3	PO4	PO5	P06	· · · · · · · · · · · · · · · · · · ·				
1	2	3	3	1	3	3	2	1	2		
2	2	3	2 1 3 2		2	2	1	2			
3	2	3	2	1	3	2	2	1	2		
4	2	3	2	1	3	2	2	1	2		
5	2	3	2	1	3	2	2	1	2		

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

	Continuo	us Assessr	nent Marks (CA		End		
Assessment	Weekly Progress	Seminar	Record work	Viva	Attendance	Semester Examination (ESE) Marks	Total Marks
Marks	40	30	10	10	10	-	100

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^{*} TE - Theory Exam, LE - Lab Exam

Department	Artificial intelligence and Data Science	Program	nme: M	.Tech.				
Semester	First	nd Semes	ter Exam	Type: -				
Course Code	P22ADC1YY	Perio	ds / We	eek	Credit	Ma	Maximum Marks	
Course Code	FZSADCIAA	L	Т	Р	С	CAM	ES	TM
Course Name	ABILITY ENHANCEMENT COURSE - I	0	0	4	-	100	<u> </u>	100
	Å	•			•			

Students shall choose an International certification course offered by the reputed organizations like Google, Microsoft, IBM, Texas Instruments, Bentley, Autodesk, Eplan and CISCO, etc. The duration of the course is 40-50 hours specified in the curriculum, which will be offered through Centre of Excellence.

Pass /Fail will be determined on the basis of participation, attendance, performance and completion of the course. If a candidate fails, he/she has to repeat the course in the subsequent years. Pass in this course is mandatory for the award of degree.

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	Artificial Intelligence and Data Scien	ce Program	me: M.	Tech.					
Semester	I	Course (Catego	ry: PE	*Enc	d Semeste	er Exam Ty	/pe: TE	
Course Code	P23ADEC01	Perio	ds / We	ek	Credit	Ma	ximum Ma	rks	
Course Coue	IZSADLOUI	L	Т	Р	С	CAM	ESE	TM	
Course Name	AGILE AND SOFTWARE PROJECT	3	3 0 0 3 40 60						
	MANAGEMENT			<u> </u>					
		(AI & DS)							
Prerequisite							DTA	1 '	
	On completion of the course, the stu	idents will be	e able t	0				Mapping est Level)	
Course	CO1 Interpret the steps involved in softwa	re developmer	 nt					K2	
Outcomes		-		:				K2	
	CO2 Identify an exclusive and appropriate design for a software project								
	CO3 Apply different software testing strategies for ensuring the quality of software							K3	
	CO4 Identify the different agile methodologies applied in the industry							K2	
	CO5 Apply different agile process used in software development						K3		
UNIT-I	Software Engineering Processes				Periods: 9				
	Fetimation - Schaduling - Rick management.	_ Software con					anagement		
Estimation Techni Requirements spe	~		figuratio	n manag	gement - Proje m structures	ect Plannin – Staffing	g – Empiric	al CO1	
Estimation Techni Requirements spe UNIT-II	iques – Staffing Level Estimation – Scheducification. Software Design	uling – Organi	figuration a	n manag and Tea	pement - Projem structures Periods: 9	ect Plannin – Staffing	g – Empiric _I – Softwar	al CO1 e	
Estimation Techni Requirements spe UNIT-II Characteristics of a Detailed Design — diagrams — State of	iques – Staffing Level Estimation – Scheducification.	uling – Organi esion – Structu Use case mod	figuration a zation a red Ana el – Cla	n manag and Tea llysis – D ss diagra	gement - Projem structures Periods: 9 Pata Flow Dia Pams – Interac	ect Plannin — Staffing grams – St	g – Empirica j – Software cructured and ms – Activity	al CO1 re	
Estimation Techni Requirements spe UNIT-II Characteristics of a Detailed Design — diagrams — State of	iques – Staffing Level Estimation – Scheducification. Software Design a Good Software Design – Coupling and Cohologic Object oriented concepts – UML Diagrams – chart diagrams – Object Oriented Analysis and	uling – Organi esion – Structu Use case mod	figuration a zation a red Ana el – Cla	n manag and Tea llysis – D ss diagra	gement - Projem structures Periods: 9 Pata Flow Dia Pams – Interac	ect Plannin — Staffing grams – Staffing grams – Staffing	g – Empirica j – Software cructured and ms – Activity	al CO1	
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Estimation Techni Requirements spe UNIT-II Characteristics of a Detailed Design — diagrams — State of Types — A User Int UNIT-III Introduction to Sof Role of a tester — S UNIT-IV Theories for Agile — Agile Manifesto	iques – Staffing Level Estimation – Scheducification. Software Design a Good Software Design – Coupling and Cohologic oriented concepts – UML Diagrams – Chart diagrams – Object Oriented Analysis and terface Design methodology. Software Testing tware testing – Psychology of Testing – Principolity of Te	esion – Structu Use case mod d Design meth ples of Softwar - Traditional Ma	red Ana el – Cla odology re Testir	and Team Ilysis – Defe	Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9	grams – Station diagra a good Use Prevention cation of A ams – Agil	g – Empirica y – Softwar cructured an ms – Activiter Interface strategies	CO1 CO2 CO3	
Estimation Techni Requirements spe UNIT-II Characteristics of a Detailed Design — diagrams — State of Types — A User Int UNIT-III Introduction to Sof Role of a tester — S UNIT-IV Theories for Agile — Agile Manifesto a Testing — Agile Do UNIT-V Lean Production	iques – Staffing Level Estimation – Scheducification. Software Design a Good Software Design – Coupling and Cohologic oriented concepts – UML Diagrams – Chart diagrams – Object Oriented Analysis and terface Design methodology. Software Testing Itware testing – Psychology of Testing – Principles of Testing – Principles – Agile Software Development – Agile Methodology Management – Agile Software Development – and Principles – Agile Project Management – Icumentations – Agile Drivers, Capabilities and	esion – Structu Use case mod d Design meth ples of Softwar Traditional Ma Agile Team Int d Values.	red Ana el – Cla odology re Testir	and Teal	Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9 Periods: 9	grams – Staffing grams – Staffing grams – Staffing grams – Staffing grams – Staffing grams – Staffing grams – Staffing	g – Empirica y – Softwar ructured an ms – Activit er Interface Strategies gile Methoc ity in Design	CO1 CO2 CO3	

Text Books

- Ian Sommerville, "Software Engineering", Pearson Education, Eighth edition, 2008.
- Craig Larman, "Agile and Iterative Development-A Manager"s Guide", Pearson Education, 2010.
- Elisabeth Hendrickson, "Agile Testing" Quality Tree Software Inc, 2012.

Reference Books

- Hazza and Dubinsky, "Agile Software Engineering, Series: Undergraduate Topics in Computer Science", Springer, 2009.
- Roger S. Pressman, "Software Engineering: A Practitioner's Approach", McGraw-Hill International Edition, Seventh Edition, 2009.
- David J. Anderson and Eli Schragenheim, "Agile Management for Software Engineering: Applying the Theory of Constraints for Business Results", Prentice Hall, 2003.
- Object-Oriented Systems Analysis and Design, McGraw-Hill Higher Education; 4th Edition, 2010.
- Robert C Martin, "Agile Software Development, Principles, Patents and Practices", Prentice Hall, 2012. 5.

Web References

- 1. https://www.coursera.org/courses?query=software%20engineering
- 2. https://www.edx.org/learn/software-engineering
- 3. https://www.udemy.com/courses/development/software-engineering/
- 4. https://www.coursera.org/learn/agile-software-development
- https://www.tutorialspoint.com/sdlc/sdlc_agile_model.htm

TE - Theory Exam, LE - Lab Exam

COs	I	Progra	m Out	Program Specific Outcomes (PSOs)					
	PO1	PO2	PO3	PO4	PO5	PO6	PSO1	PSO2	PSO3
1	2	3	2	-	-	-	1	2	1
2	3	3	3	-	1	-	2	2	1
3	2	3	2	1	2	•	2	2	1
4	3	3	3	1	3	1	2	2	1
5	3	3	3	1	3	-	2	2	1

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus **Assignment to be given from Unit-5

5. X } \ __

	Artificial Intelligence and Data Science	Program	nme: M.	Tech.				
Semester	I	Course	Categor	ry: PE	*End	d Semeste	er Exam Ty	/pe: TE
Course Code	P23ADE101	Perio	ds / We	ek	Credit	Ma	ximum Ma	rks
Course Code	FZSADLIUI	L	Т	Р	С	CAM	ESE	TM
Course Name	PYTHON FOR DATA SCIENCE	3	0	0	3	40	60	100
Prerequisite	Python Basics			į				
•	On completion of the course, the studer	nts will be	e able t	0				Mapping est Level)
Course	CO1 Identify the roles and stages of data scie	nce projec	ts					K2
Outcomes	CO2 Interpret data structures provided by pan	das and n	umpy lib	rary for o	lata analysis			K2
	CO3 Perform data wrangling, cleaning and tra	nsformatio	n using	python				K2
	CO4 Apply matplotlib for plotting and visualizing	ng the data	asets					K3
	CO5 Apply data aggregation and time series a	analysis us	ing pytho	on progra	ımming			K3
UNIT-I	Introduction to Data Science				Periods: 9			
	ess - Roles, Stages in data science project - Wo			11 11169 –	WORKING WILL	i i Cialionai	ualabases	٦
UNIT-II	lanaging data – Cleaning and sampling for modeli Basics of Numpy, Pandas, and Vectorize	ed Comp	utation		Periods: 9			
UNIT-II The Numpy ndarra Using Arrays - File pandas Data Strue	Basics of Numpy, Pandas, and Vectorize ay: A Multidimensional Array Object – Universal Full a Input and Output with Arrays – Linear Algebra – betures – Essential Functionality – Summarizing an	ed Comp unctions: F Random I	utation ast Elen Number	nent-wise Generati	Array Functon – Randon	tions – Dat n Walks. Ir	troduction i	ng CO 2
UNIT-II The Numpy ndarra Using Arrays - File	Basics of Numpy, Pandas, and Vectorize ay: A Multidimensional Array Object – Universal Full a Input and Output with Arrays – Linear Algebra – betures – Essential Functionality – Summarizing an	ed Comp unctions: F Random I nd Compu	utation Fast Elen Number ting Des	nent-wise Generati	Array Functon – Randon	tions – Dat n Walks. Ir landling Mi	troduction i	to
UNIT-II The Numpy ndarra Using Arrays - File pandas Data Struc Hierarchical Index UNIT-III Data preprocessin Interacting with Da Data wrangling an	Basics of Numpy, Pandas, and Vectorize ay: A Multidimensional Array Object – Universal Form Input and Output with Arrays – Linear Algebra – Cures – Essential Functionality – Summarizing and Data Preprocessing, Wrangling, and Tra g: Reading and Writing Data in Text Format – Estabases. d transformation: Combining and Merging Data S	ed Compunctions: F Random I nd Compunctions Insforma	utation Fast Elen Number ting Des tion a Format	nent-wise Generati criptive S	Array Function – Randon Statistics – H Periods: 9 acting with I	tions – Dat n Walks. Ir landling Mi HTML and	ssing Data Web APIs	ng CO2
UNIT-II The Numpy ndarra Using Arrays - File pandas Data Struc Hierarchical Index UNIT-III Data preprocessin Interacting with Da Data wrangling an	Basics of Numpy, Pandas, and Vectorize ay: A Multidimensional Array Object – Universal For Enput and Output with Arrays – Linear Algebra – Ctures – Essential Functionality – Summarizing along. Data Preprocessing, Wrangling, and Trage: Reading and Writing Data in Text Format – Entabases.	ed Compunctions: F Random I nd Compunctions Insforma	utation Fast Elen Number ting Des tion a Format	nent-wise Generati criptive S	Array Function – Randon Statistics – H Periods: 9 acting with I	tions – Dat n Walks. Ir landling Mi HTML and Transforma	ssing Data Web APIs	ng CO2
UNIT-II The Numpy ndarra Using Arrays - File pandas Data Struc Hierarchical Index UNIT-III Data preprocessin Interacting with Da Data wrangling an Manipulation – US UNIT-IV A Brief matplot lib	Basics of Numpy, Pandas, and Vectorize ay: A Multidimensional Array Object – Universal Fultures – Essential Functionality – Summarizing and Ing. Data Preprocessing, Wrangling, and Trage: Reading and Writing Data in Text Format – Estabases. d transformation: Combining and Merging Data SDA Food Database Plotting and Visualization API Primer – Plotting Functions in pandas – Plotting and Plottons in pandas – Plotting Plottons in pandas – Plotting Plottons in pandas – Plotting Plottons in pandas – Plotting Plottons in pandas – Plotting Plotting Plottons in pandas – Plotting Plottin	ed Compunctions: F Random I and Compunctions Insformations Binary Data	utation Fast Elen Number ting Des tion a Format	nent-wise Generati criptive S ts – Inter and Pivo	Array Function – Randon Statistics – H Periods: 9 acting with I ting – Data	tions – Dat n Walks. Ir landling Mi HTML and Transforma	ssing Data Web APIs	ng CO:
UNIT-II The Numpy ndarra Using Arrays - File pandas Data Structure Hierarchical Index UNIT-III Data preprocessin Interacting with Data Data wrangling an Manipulation – US UNIT-IV A Brief matplot lib	Basics of Numpy, Pandas, and Vectorize ay: A Multidimensional Array Object – Universal Fultures – Essential Functionality – Summarizing and Ing. Data Preprocessing, Wrangling, and Trage: Reading and Writing Data in Text Format – Estabases. d transformation: Combining and Merging Data SDA Food Database Plotting and Visualization API Primer – Plotting Functions in pandas – Plotting and Plottons in pandas – Plotting Plottons in pandas – Plotting Plottons in pandas – Plotting Plottons in pandas – Plotting Plottons in pandas – Plotting Plotting Plottons in pandas – Plotting Plottin	ed Compunctions: F Random I and Compunctions Insformations Binary Data Sets – Res	utation Fast Elen Number ting Des tion a Format shaping	nent-wise Generatic criptive S ts – Inter and Pivo	Array Function – Randon Statistics – H Periods: 9 acting with I ting – Data	tions – Dat n Walks. Ir landling Mi HTML and Transforma	ssing Data Web APIs	og CO:
UNIT-II The Numpy ndarra Using Arrays - File pandas Data Struct Hierarchical Index UNIT-III Data preprocessin Interacting with Data wrangling an Manipulation – US UNIT-IV A Brief matplot lib Visualization Tool UNIT-V Data aggregation of Pivot Tables and C Time series: Date Handling – Period	Basics of Numpy, Pandas, and Vectorize ay: A Multidimensional Array Object – Universal Fulter and Output with Arrays – Linear Algebra – Edures – Essential Functionality – Summarizing and Ing. Data Preprocessing, Wrangling, and Trage: Reading and Writing Data in Text Format – Estabases. d transformation: Combining and Merging Data States DA Food Database Plotting and Visualization API Primer – Plotting Functions in pandas – Plotecosystem. Data Aggregation, Group Operations and and group operations: GroupBy Mechanics – Data	ed Compunctions: F Random I nd Compu Insforma Binary Data Sets – Res Otting Mapu d Time S a Aggrega Basics – D	utation Fast Elen Number ting Des tion a Format shaping s: Visual series tion – G Date Ran	nent-wise Generation Criptive S ts – Inter and Pivo lizing Ha roup-wise ges, Free	Periods: 9 tit Earthquak Periods: 9 tacting with I ting – Data Periods: 9 tit Earthquak Periods: 9 tit Earthquak Periods: 9 tit Earthquak	tions – Date on Walks. In landling Mi HTML and Transforma te Crisis Date and Trans and Shifting	web APIs ation – Strir ata – Pytho	on CO4

- 2. Jake VanderPlas, "Python Data Science Handbook Essential tools for working with data", O'Reilly, 1st edition, 2016.
- 3. Fabio Nelli, "Python Data Analytics with Pandas, NumPy, and Matplotlib", 2nd edition, 2018.

Reference Books

- 1. John Paul Mueller, Luca Massaron, "Python for Data Science for Dummies", John Wiley& Sons, 2nd edition, 2019.
- 2. Jesus Rogel-Salazar, "Data Science and Analytics with Python", CRC Press Taylor and Francis Group, 1st edition, 2017.
- 3. Mark Lutz, Laura Lewin, Frank Willison, "Programming Python", O'Reilly Media, 3rd edition, 2006.
- 4. Eric Matthes, "Python Crash Course: A Hands-on, Project-based Introduction to Programming", 2nd edition, 2019
- 5. Al Sweigart, "Automate the Boring Stuff with Python: Practical Programming for Total Beginners", 1st edition, 2015

Web References

- 1. https://nptel.ac.in/courses/106/106/106106212/
- https://www.geeksforgeeks.org/data-analysis-visualization-python/
- https://www.coursera.org/learn/python-data-analysis
- https://www.python.org/
- 5. https://www.datacamp.com/courses/statistical-thinking-in-python-part-1

COs	!	Progra	m Out		ram Spe omes (P				
	PO1	PO2	PO3	PO4	PO5	PO6	PSO1	PSO3	
1	2	2	2	1	2	2	3	2	2
2	1	2	2	2	2	2	3	2	2
3	2	3	3	1	3	3	3	3	3
4	2	3	3	1	3	3	3	3	3
5	2	3	3	1	3	3	3	3	3

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Methods

Assessment		Continu	ious Asse	essment Marks(CAM)	End Semester	Total
	CAT 1	CAT 2	Model	Assignment*	Attendance	Examination(ESE)	Marks
			Exam			Marks	
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus **Assignment to be given from Unit-5

J. X } __

Department	Artific	ial Intelligence and Data Science	Program	me: M.	Tech.				
Semester	I		Course	Catego	ry: PE	*End	d Semeste	r Exam Ty	/pe: TE
Course Code	Ρ23ΔΙ	DE102	Perio	ds / We	ek	Credit	Ma	ximum Ma	rks
Course Code	1 23/1	JE 102	L	Т	Р	С	CAM	ESE	TM
Course Name	DATA	SCIENCE ESSENTIALS	3	0	0	3	40	60	100
	<u> </u>	(A	AI & DS)				<u>.</u>		
Prerequisite	-								
	On co	mpletion of the course, the studer	nts will be	e able t	0				/lapping est Level
Course	CO1	Interpret the data science process and	how its co	mponen	ts intera	act			K2
Outcomes	CO2	Classify, formulate the data science pro	oblems an	d manag	ge large	dataset			K2
	CO3	Implement modeling, apply basic mach	nine learnir	ng algori	thms an	nd evaluate the	e model.		K3
	CO4	Apply effective visualization and text a	nalysis for	data sci	ence pr	ojects			K3
	CO5	Apply data science toolkit and develop	awareness	of ethic	al dime	nsions of data	science		K3
UNIT-I	Introd	uction				Periods: 9			
Scraping the Web Dimensionality Re	Dbjective - Using eduction	Management s - Datasets - Descriptive Statistics - Dat APIs – Working with Data: Exploring You – Data Handling: The problems of handl ributing data storage and processing witl	ur Data - C ling large c	leaning a lata – Te	and Mur	nging - Manipu es and Progra	lating Data mming tips	– Rescalin	g
UNIT-III		Modeling and Algorithms				Periods: 9			
Igorithms: Classif	ication: k	achine learning in Data Science - Overf -Nearest Neighbors - Naïve Bayes - Sup stering: Kmeans Clustering – Model Eval	port Vecto	r Machir	nes– Re	gression Meth	nods: Linea	r Regressio	
UNIT-IV	Data \	/isualization and Text Analysis				Periods: 9			
Data Types - Data	ata Enco eh (Pytho	Data Visualization History - Types of Data odings - Retinal variables - Mapping v on) – Text mining and Text Analytics: Tex	ariables to	Encod	ings - 🕽	Visual encodi	ngs - Tech	nologies fo	or
UNIT-V	:	Science Tools and Ethics				Periods: 9			<u>i</u>
Tools - Data Visua	alization	duce, Pregel, and Hadoop – RapidMiner: Tools - Data Transformation Tools - Sam Data Scientists, Hubris, and Ethics.							
Lecture Period		Tutorial Periods: 15	Practica	al Perio	ds: -	Т	otal Perio	ds: 45	
ext Books						·			
		a Deshpande, Data Science, Concepts a							

- Davy Cielen, Arno D. B. Meysman and Mohamed Ali, Introducing Data Science: Big Data, Machine Learning, and more, using Python Tools, Manning, 2016.
- 3. Cathy O"Neil and Rachel Schutt, Doing Data Science, Straight Talk from The Frontline, O"Reilly, 2013

Reference Books

- 1. Joel Grus, Data Science from Scratch, Second Edition, O"Reilly, 2019.
- 2. Skiena, Steven S.. The Data Science Design Manual., Springer, 2017.
- 3. Foster Provost and Tom Fawcett, Data Science for Business: What You Need to Know About Data Mining and Data- Analytic Thinking, 1st edition, 2013.
- 4. John Paul Mueller and Luca Massaron, Python for Data Science for Dummies, 1st edition, 2015.
- 5. Christopher M. Bishop, Pattern Recognition and Machine Learning, 1st edition, 2006.

Web References

- 1. https://www.ibm.com/topics/data-science
- 2. https://www.oracle.com/in/what-is-data-science/
- 3. https://u-next.com/blogs/data-science/importance-of-data-science/
- 4. https://monkeylearn.com/text-analysis/
- 5. https://www.toptal.com/designers/data-visualization/data-visualization-tools

TE - Meory Exam, LE -

* TE – Theory Exam, LE – Lab Exam

COs	1	Progra	m Out	comes	(POs))		ram Spe omes (P	
	PO1	PO2	PO3	PO4	PO5	PO6	PSO1	PSO2	PSO3
1	2	3	2	-	-	-	1	2	1
2	3	3	3	-	1	-	2	2	1
3	2	3	2	1	2	•	2	2	1
4	3	3	3	1	3	ı	2	2	1
5	3	3	3	1	3	-	2	2	1

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus **Assignment to be given from Unit-5

Department	Artific	ial Intel	ligence and Data Science	Program	me: M.	Tech.				
Semester	I			Course (Catego	ry: PE	*E	nd Semester	Exam Typ	e: TE
Course Code	P23A[7F103		Perio	ds / We	ek	Credit	Maxi	mum Marl	(S
Course Code	I ZUAL	JE 103		L	Т	Р	С	CAM	ESE	TM
Course Name	BIG D	ATA MII	NING AND ANALYTICS	3	0	0	3	40	60	100
			(/	AI & DS)			•			
Prerequisite	-									
	On co	mpletio	n of the course, the stude	nts will be	e able t	0			1	apping it Level)
1 1	CO1	Interpre	et the fundamental concepts of	big data an	d analyt	ics.			K	2
Outcomes	CO2	Apply F	ladoop and map-reduce techni	ques for big	g data a	pplication	ns		K	3
	CO3	Apply a	algorithms for handling petabyte	s of datase	ets				K	2
	CO4	Apply consum	algorithms and propose solunption	utions for	Big Da	ta by	optimizing	main memory	′ K	3
	CO5	Apply pr	redictive analysis in real time ap	oplications					K	3
UNIT-I	Introd	uction 1	To Big Data and Analytics				Periods:	9		
	tics role	of data	mportance of Big data – Big da scientists - Key roles for succe: ytics applications							
,	HADO						Periods:	9		
Hadoop Streaming-	Design a Map I	of HDFS Reduce J	uted File System – Component S-Java interfaces to HDFS Bas lob runFailures-Job Schedulino doop environment.	ics - Devel	oping a	Map Re	educe Appli	cation- How M	ap Reduce	
UNIT-III	Simila	r Items	Search				Periods:	9		
			g of Documents – Similarity pre lity Sensitive Functions – LSH						ocuments –	СОЗ
UNIT-IV	Mining	g Data S	Streams				Periods:	9		
			a in the Stream – Filtering Str w – Decaying Windows	eams – Co	ounting I	Distance	e Elements	in a Stream –	Estimating	CO4
UNIT-V	Predic	tive An	alytics				Periods:	9		
			egression- Multiple linear regre eraction techniques - Systems			ion of re	egression co	efficients. Visu	ıalizations -	CO5
Lecture Periods	: 30		Tutorial Periods: 15	Practica	l Perio	ds: -		Total Period	ls: 45	
				4						

Text Books

- Chris Eaton, Dirk DeRoos, Tom Deutsch, George Lapis, Paul Zikopoulos, "Understanding Big Data: Analytics for Enterprise Class Hadoop and Streaming Data", McGraw-Hill Publishing, 2012.
- 2. ure Leskovec, Anand Rajaraman, Jeffrey David Ullman, "Mining of Massive Datasets", Cambridge University Press, 2nd Edition, 2014.
- 3. Jiawei Han, Micheline amber, Jian Pei, "Data Mining Concepts and Techniques", Morgan Kaufman Publications, 3rd Edition, 2011.

Reference Books

- Ian H.Witten, Eibe Frank "Data Mining Practical Machine Learning Tools and Techniques", Morgan Kaufman Publications, 4th Edition, 2016.
- 2. Arshdeep Bahga, Vijay Madisetti, "Big Data Science & Analytics: A Hands-On Approach ", VPT, 1st Edition, 2018.
- 3. Michael Minelli, Michehe Chambers, "Big Data, Big Analytics: Emerging Business Intelligence and Analytic Trends for Today's Business", 1st Edition, Ambiga Dhiraj, Wiely CIO Series, 2013.
- Viktor Mayer-Schönberger and Kenneth Cukier, Big Data: A Revolution That Will Transform How We Live, Work, and Think, 1st edition, 2013.
- 5. Tom White, Hadoop: The Definitive Guide, 4th edition, 2015.

Web References

- 1. https://encyclopedia.pub/entry/12788#:~:text=Big%20data%20mining%20(BDM)%20is,data%20of%20an%20immense%20volume.
- 2. https://www.techopedia.com/definition/30215/big-data-mining
- 3. https://www.techtarget.com/searchbusinessanalytics/definition/data-mining
- 4. https://www.javatpoint.com/types-of-sources-of-data-in-data-mining-in-dbms
- 5. https://www.ibm.com/topics/predictive-analytics

* TE - Theory Exam, LE - Lab Exam

2. 1

COs	İ	Progra	m Out	comes	(POs)		ram Spe omes (P	PSOs)	
	PO1	PO2	PO3	PO4	PO5	P06	PSO1	PSO2	PSO3	
1	2	3	2	-	-	-	1	2	1	
2	3	3	3	-	1	-	2	2	1	
3	2	3	2	1	2	1	2	2	1	
4	3	3	3	1	3	-	2	2	1	
5	3	3	3	1	3	-	2	2	1	

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus
**Assignment to be given from Unit-5

J. X } \ _

Department	Artifi	cial Intelligence and Data Science	Progran	nme: M.	Tech.				
Semester	I		Course	Categoi	y: PE	*End	Semeste	r Exam Ty	pe: TE
Course Code	DOSA	DE104	Perio	ds / We	ek	Credit	Max	kimum Ma	rks
Course Code	FZJA	DE 104	L	Т	Р	С	CAM	ESE	TM
Course Name	1	FICIAL INTELLIGENCE FOR SION MAKING	3	0	0	3	40	60	100
		(A	AI & DS)						
Prerequisite	-								
	On co	ompletion of the course, the studer						(Highe	lapping st Level)
Course	CO1	Interpret the concepts behind the expert	systems a	nd apply	it in cre	ating rule-bas	sed systems	3.	K2
Outcomes	CO2	Identify different knowledge representation	on techniq	ues.					K2
	CO3	Apply various Inference methods for reas	soning						K3
	CO4	Apply probabilistic methods for performing	ng reasoni	ng under	uncerta	inty.			K3
	CO5	Apply the concepts for creating expert sy	stem in re	al time					K3
UNIT-I		luction to Expert Systems				Periods: 9			
n the developmen	it of an e	system - problem domain and knowledge expert system - general characteristics of - procedural and nonprocedural paradign	an expert	system	- history	and uses of	expert syst		
UNIT-II	The I	Representation of Knowledge				Periods: 9			
imitations of prop UNIT-III	ositional Metho	symbols to represent knowledge - the mean and predicate logic. Pods of Inference Pohs - state and problem spaces - AND-O				Periods: 9			
imitations of propo easoning - applyir knowledge - the Ma	sitional l ng resolu arkov de	logic - logic systems - resolution rule of infection to first-order predicate logic - forward cision process – Decision Making – Decisiecision Support System.	erence - re l and back	solution ward cha	systems ining - a	s - and deduct dditional meth	ion - shallov nods of Infe	v and causa rence - Met	al CO: a
UNIT-IV		ning Under Uncertainty				Periods: 9			
nduction - feature hypothetical reaso	ncertaint s of clas ning and	y and theories devised to deal with it - to ssical probability - experimental and sub d backward induction - temporal reasoning ace chains - implications of combining evidence	ojective pro ng - Marko	babilities v chains	s - com _l - odds -	oound and co of belief - suf	onditional p ficiency and	robabilities d necessity	-
UNIT-V	Desig	n of Expert Systems				Periods: 9			
	ges - the	oriate problem - the stages in the develor e role of the knowledge engineer in the b cycle model.							
Lecture Period		Tutorial Periods: 15	Practic	al Perio	ds: -	Т	otal Perio	ds: 45	
ext Books									
	. ,	tems Design and Development", Macmilla	•						
	•	ng Expert Systems", West Publishing Cor							
		uction to Expert Systems", Addison Wesle	ey Longma	an, 1999.					
Reference Book		val "The Engineering of Knowledge De-	d Cyctor-	" D+	oo Llall	1004			
 Nikolopoulos, H. B. Verbrug Lakhmi C. Jai 	"Expert gen, Spy n, Gloria	kel, "The Engineering of Knowledge-Base Systems", Marcel Dekker Inc. 1997. rros G. Tzafestas, "Artificial Intelligence in Phillips-Wren, "Intelligent Decision Supp Kumar Rout, Himansu Das, Suresh Cha	n Industrial ort System	Decisior s in Age	n Making nt-media	g, Control and ated Environn	nents", IOS	Press, 200	5.

5. Nilanjan Dey, Jitendra Kumar Rout, Himansu Das, Suresh Chandra Moharana "Applied Intelligent Decision Making in Machine Learning", CRC Press; 1st Edition, 2020.

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Web References

- 1. https://ocw.mit.edu/courses/electrical-engineering-and-computer-science/6-034-artificial-intelligence-fall-2010/lecture-videos/lecture-3-reasoning-goal-trees-and-rule-based-expert-systems/
- 2. http://www.umsl.edu/~joshik/msis480/chapt11.htm
- 3. https://www.coursera.org/courses?query=decision%20making
- 4. https://www.slideshare.net/akhilrocker143/572-11293384
- 5. https://www.sciencedirect.com/science/article/abs/pii/0378720693900696

COs/POs/PSOs Mapping

COs	I	Progra	m Out	comes	(POs			ram Spe omes (P	
	PO1	PO2	PO3	PO4	PO5	PO6	PSO1	PSO2	PSO3
1	2	3	2	-	-	-	1	2	1
2	3	3	3	-	1	-	2	2	1
3	2	3	2	1	2	-	2	2	1
4	3	3	3	1	3	-	2	2	1
5	3	3	3	1	3	-	2	2	1

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

J. N. S.

^{*} TE - Theory Exam, LE - Lab Exam

^{**}Assignment to be given from Unit-5

Department	Artific	ial Intelligence and Data Scienc	e Progran	nme: M	.Tech.				
Semester	II		Course	Catego	ry: PC	*Eı	nd Semeste	r Exam ⁻	Гуре: ТЕ
Course Code			Perio	ods/We	ek	Credit	Max	imum Ma	arks
	P23A[L	Т	Р	С	CAM	ESE	TM
Course Name		LLEL PROGRAMMING DIGMS	3	0	0	3	40	BT Mapp (Highest L K3 K3 K3 K3 C3 K3 K3 C3 C3 C3 C3 C4 C5 C5 C5 C5 C5 C5 C5 C5 C5 C5 C5 C5 C5	100
		(AI&DS)							
Prerequisite	Oper	ating System Concepts, Compute	er Architecture	e and O	rganiza	tion			
		f the course, the students will b						(Highes	st Level
Course Outcome	CO1	Apply MPI framework for passing me		-	-	sses		K	(3
Outcome	CO2	Apply Pthreads for creating shared r	memory paralle	el progra	ıms			K	(3
	CO3	Apply OpenMP paradigms to create	shared memo	ry parall	el progra	ıms		K	(3
	CO4	Apply either OpenMP, MPI for parall	lel algorithms f	or searc	hing and	sorting		k	(3
	CO5	Apply CUDA programming for config	guring hardwar	e and tra	ansfer da	ata across GF	PU and CPU	k	(3
UNIT – I	Mess	age Passing Paradigm				Periods:9			
completion									
and MPI_Recv –	Message Illgather	 MPI_Init and MPI_Finalize - MPI or matching - MPI I/O - Parallel I/O - Or matching - MPI I/O - Remote Memory Area Memory Paradigm: pthreads 	Collective comr ccess – Perfor	nunicatio	on – MPI	_Reduce - M	PI_Allreduce grams		
		read synchronization – Critical section		ting _ M	lutov – 9			nd condition	n CO2
		cks with examples - Caches, cache co						ia conditio	JII CO2
UNIT – III	Shar	ed Memory Paradigm: openMP				Periods:9			<u> </u>
Basic OpenMP c	onstruct	s – scope of variables – Reduction cl	ause – Paralle	l For dire	ective –	loops in Ope	nMP – Sche	duling loop	os
pest practices	n in Ope	nMP – Case Study: Producer-Consur	mer problem –	Cache is	ssues – ¯	Γhreads safet	y in OpenMF	OpenM	^{1P} CO3
UNIT – IV	1	llel Algorithms				Periods:9			
		thms: Reduction – Broadcast - Prefix nulticomputer. Sorting: Odd even tran						- Algorith	m CO4
UNIT – V	••••	Programming with CUDA	isposition sort	Ditorne	merge -	Periods:9	~		L
GPUs and GPGF	PU - GPL	J architectures - Heterogeneous com	puting – Simple	e CUDA	program	ı - Threads, b	locks, and g	ids - Vect	or
addition – CUDA nore than one w	trapezo	oidal rule – improvements - Implement	ntation of trape	ezoidal r	ule with	warpSize thr	ead blocks -	- block wi	th CO5
LecturePerio		TutorialPeriods:0	Practic	al Perio	ods0	7	TotalPeriod	ls:45	
Text Books	u3.73	rutorian crious.o	Tractio	ar i cric	Jus0		otali cilot	13.43	
	. Pached	o, Matthew Malensek, "An introduction	on to parallel pr	ogramm	nina". Se	cond edition.	Morgan Kau	fmann, 20)21
		lunkar, Raju K, "Introduction to Parall	•					, -	
3. Michael	J. Quinr	n, "Parallel Computing: Theory & Prac	ctice", Tata Mc	Graw Hil	II. Secon	d edition, Re	orint 2017.		
		i, Faranei Compuning. Theory & Frac			.,				
		i, Farallel Computing. Theory & Frac			,				
Reference Boo	oks	aster, T. G. Mattson, J. Fung, and D.	Ginsburg, "Op	enCL pro		ing guide", Ad	ldison Wesle	y, 2011	
Reference Boo 1. A. Muns	o ks shi, B. Ga			-	ogramm		ldison Wesle	y, 2011	
1. A. Muns 2. M. J. Q	oks shi, B. Ga uinn, "Pa	aster, T. G. Mattson, J. Fung, and D.	l OpenMP", Ta	ta McGr	ogramm aw Hill, 2		ddison Wesle	y, 2011	
Reference Boo 1. A. Muns 2. M. J. Qu	o ks shi, B. Ga uinn, "Pa rber, "CU	aster, T. G. Mattson, J. Fung, and D. rallel programming in C with MPI and	l OpenMP", Ta	ta McGr	ogramm aw Hill, 2		ddison Wesle	y, 2011	
1. A. Muns 2. M. J. Q. 3. Rob Far	oks shi, B. Ga uinn, "Pa rber, "CU es	aster, T. G. Mattson, J. Fung, and D. rallel programming in C with MPI and	l OpenMP", Ta lent", Morgan I	ta McGr	ogramm aw Hill, 2		ddison Wesle	у, 2011	
1. A. Muns 2. M. J. Qi 3. Rob Fai Web Reference 1. http	oks shi, B. Ga uinn, "Pa rber, "CU es o://condo	aster, T. G. Mattson, J. Fung, and D. rallel programming in C with MPI and IDA application design and developm	OpenMP", Ta ent", Morgan F	ta McGr	ogramm aw Hill, 2		ddison Wesle	y, 2011	
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A. Muns 2. M. J. Q. 3. Rob Fai Web Reference 1. http 2. http 3. http	oks shi, B. Ga uinn, "Pa rber, "CU es o://condo o://www.l	aster, T. G. Mattson, J. Fung, and D. rallel programming in C with MPI and IDA application design and developm r.cc.ku.edu/~grobe/docs/intro-MPI-C.npcc.unn.ru/mskurs/ENG/DOC/pp09.	I OpenMP", Ta lent", Morgan I shtml	ta McGra	ogramm aw Hill, :		ddison Wesle	y, 2011	

^{*} TE - Theory Exam, LE - Lab Exam

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COs	I	Progra	m Out	comes	(POs))		ram Spe omes (P	
	PO1	PO2	PO3	PO4	PO5	P06	PSO1	PSO2	PSO3
1	2	3	2	-	-	-	1	2	1
2	3	3	3	-	1	-	2	2	1
3	2	3	2	1	2	-	2	2	1
4	3	3	3	1	3	-	2	2	1
5	3	3	3	1	3	-	2	2	1

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus
**Assignment to be given from Unit-5

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Department	Artific	ial Intelligence and Data Science	Progran						
Semester	II		Course	Categor	y: PC	*Enc	l Semeste	er Exam Typ	e: TE
Course Code	Ρ23ΔΙ	OT205	Perio	ds / We	ek	Credit	Ma	ximum Mark	S
Course oode	1 204	31200	L	Т	Р	С	CAM	ESE	TM
Course Name	NATU	RAL LANGUAGE PROCESSING	3	0	0	3	40	60	100
		(/	AI & DS)		•		·		
Prerequisite	Machi	ne Learning							
	On co	mpletion of the course, the studer						BT Ma (Highes	
Course	CO1	Interpret the basics of NLP for process	sing word,	oaragrap	h and s	entence		K	3
Outcomes	CO2	Identify and apply the basic ML and DI	_ techniqu	es for NL	Р			K2,	K3
	CO3	Interpret and realize the advanced dee	p learning	paradigr	ns along	NLP Techni	ques.	K2,	K3
	CO4	Identify the concept of NLU, NLG and	apply the	concept c	of Inform	ation Retrieva	al	K2,	K3
	CO5	Apply ethics and NLP Libraries to be fo	llowed wh	le buildin	g NLP A	Applications		K	3
UNIT – I	Introd	luction				Periods:9		L	
	ext Prep	rocessing: Tokenization, Stemming and		ation, Po	s Taggi	ng, Named E	ntity Reco	gnition. NLP	
.	··••	Count Vector, Word Sense Disambiguat	ion						CO1
JNIT – II		uage Modelling				Periods:9			
		arkov Models, Maximum Likelihood Estima atiment Analysis, Topic Modelling and Cl							CO2
UNIT – III	Advar	nced NLP Techniques				Periods:9			
Sequence- to -Sec	quence M	odels, Attention Mechanisms, Transform	er Archite	cture: BE	RT, GP	Γ			CO3
UNIT – IV	Langu Retrie	uage Understanding and Generational contractions and Generation (Contraction) and Generation (Contracti	on, Inforn	nation		Periods:9			
ndexing and Sear		Answering, Dialogue Systems and Chatb	iolo, iviaci ii	ne mans	nation, C	1055 Liliquai	Hallslei L	eammu. Text	
LINIT — V								3	
UNIT – V Bias and Fairness	NLP T	ools, Libraries, Applications, Ethi	cs			Periods:9			
Bias and Fairness	NLP T	Tools, Libraries, Applications, Ethio P, Privacy Concerns in NLP Application alysis, Named Entity Recognition in Real	cs ns. NP lib World Da	raries: a Sets, 1	NLTK, S	Periods:9 Spacy, Tens	or Flow, F	Pytorch. NLP	
Bias and Fairness Applications: Senti LecturePeriod	NLP T s in NLF ment An	Tools, Libraries, Applications, Ethion, Privacy Concerns in NLP Application	cs ns. NP lib	raries: a Sets, 1	NLTK, S	Periods:9 Spacy, Tensessification for	or Flow, F	Pytorch. NLP omains.	
Bias and Fairness Applications: Senti LecturePeriod Fextbooks	NLP T s in NLF ment And s:45	Tools, Libraries, Applications, Ethion, Privacy Concerns in NLP Application alysis, Named Entity Recognition in Real	cs ns. NP lib World Da	raries: a Sets, 1	NLTK, S	Periods:9 Spacy, Tensor Spacy, Tensor Spacy, Tensor Space Sp	or Flow, F Various Do ecturePe	Pytorch. NLP omains.	
Bias and Fairness Applications: Senti LecturePeriods Textbooks 1. Vijay Mad	NLP Ts in NLF iment Ans	Tools, Libraries, Applications, Ethio P, Privacy Concerns in NLP Application alysis, Named Entity Recognition in Real TutorialPeriods:0	cs ns. NP lib World Da Practic a A Hands-0	raries: a Sets, 1 alPeriocon	NLTK, S Fext Class Is:0	Periods:9 Spacy, Tenses sification for L PT, 1st editio	or Flow, F Various Do ecturePe	Pytorch. NLP omains.	
Bias and Fairness Applications: Senti LecturePeriod: Textbooks 1. Vijay Mac 2. James Al	NLP 1 s in NLF iment Ans s:45 disetti ans	Tools, Libraries, Applications, Ethiopolications, Privacy Concerns in NLP Applicationallysis, Named Entity Recognition in Real TutorialPeriods:0 d Arshdeep Bahga, "Internet of Things: ural Language Understanding", 2nd Editional Edition	cs ns. NP lib World Da Practica A Hands-on, Pearson	raries: a Sets, 1 alPerioc on Appro	NLTK, Sext Class Is:0 ach", VF	Periods:9 Spacy, Tenses sification for L PT, 1st editio 3.	or Flow, F Various Do ecturePe n,2014.	Pytorch. NLP omains.	
Bias and Fairness Applications: Senti LecturePeriod: Textbooks 1. Vijay Mac 2. James Al 3. Jurafsky,	NLP To some in NLF iment And s:45 disetti and len, "Nation Dan and	Tools, Libraries, Applications, Ethion, Privacy Concerns in NLP Application alysis, Named Entity Recognition in Real TutorialPeriods:0 d Arshdeep Bahga, "Internet of Things: ural Language Understanding", 2nd Edition Martin, James, "Speech and Language	cs ns. NP lib World Da Practica A Hands-on, Pearson Processing	raries: a Sets, 1 alPeriod on Appro	NLTK, Sext Classification (No. 1) NLTK, Sext Classification, Vision, 2003	Periods:9 Spacy, Tensissification for L PT, 1st edition 3. rentice Hall, 2	or Flow, F Various Do ecturePe n,2014.	Pytorch. NLP omains.	COS
Bias and Fairness Applications: Senti LecturePeriod: Textbooks 1. Vijay Mac 2. James Al 3. Jurafsky, 4. Srini Jana	NLP 1 s in NLF iment Ans s:45 disetti and len, "Nati Dan and arthanam	Tools, Libraries, Applications, Ethiopolications, Privacy Concerns in NLP Applicationallysis, Named Entity Recognition in Real TutorialPeriods:0 d Arshdeep Bahga, "Internet of Things: ural Language Understanding", 2nd Editional Edition	cs ns. NP lib World Da Practica A Hands-on, Pearson Processing	raries: a Sets, 1 alPeriod on Appro	NLTK, Sext Classification (No. 1) NLTK, Sext Classification, Vision, 2003	Periods:9 Spacy, Tensissification for L PT, 1st edition 3. rentice Hall, 2	or Flow, F Various Do ecturePe n,2014.	Pytorch. NLP omains.	CO5
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* TE - Theory Exam, LE - Lab Exam

COs	l	Progra	m Out	Program Specific Outcomes (PSOs)					
	P01	PO2	PO3	PO4	PO5	P06	PSO1	PSO2	PSO3
1	2	3	2	-	-	-	1	2	1
2	3	3	3	-	1	1	2	2	1
3	2	3	2	1	2	1	2	2	1
4	3	3	3	1	3	-	2	2	1
5	3	3	3	1	3	-	2	2	1

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

5. X } \ __

^{**}Assignment to be given from Unit-5

Department	Artific	ial Intelligence and Data Science	Program	nme: M	.Tech.				
Semester	II		Course	Catego	ry: PC	*Enc	d Semeste	r Exam T	ype: TE
Course Code	P23A1	DT206	Perio	ds / We	eek	Credit	Ma	ximum Ma	arks
Course Code	1 23/1	31200	L	T	Р	С	CAM	ESE	TM
Course Name	ADVA	NCED DEEP LEARNING	3	0	0	3	40	60	100
		(/	AI & DS)		ii.		i	i	
Prerequisite	Machi	ne Learning Algorithms							
	On co	mpletion of the course, the studer	nts will be	able t	to			BT N	Mapping
_									est Level)
Course Outcomes	CO1	Identify various neural network and ac accuracy	tivation fun	ction to	calculate	e the loss and	l improve ti	ne	K2
Outcomes	CO2	Apply different Convolutional Neural N	etwork for	process	ina imaa	es effectively			K3
	CO3	Interpret and apply deep learning regu		•				K	2, K3
	CO4	Apply different Neural Network Models					ction etc		K3
	CO5		•					a d	K3
	COS	Apply Neural Style transfer and autoer content	icoding pro	ocess ic	or illialing	duplicates in	i images ai	IG	N3
UNIT – I	Found	dations Of Neural Networks				Periods:9			
		logical Neuron-The Perceptron - Multi	layer Feed	l - Forv	ward Net		ining Neur	al Network	s:
Backpropagation Le	earning -	· Activation Functions: Linear – Sigmoid -	- Ťanh - Ha	rd Tanh	- Softm	ax -Rectified	Linear - Los	ss Function	ns: CO1
		oss Functions for Regression - Loss Fu Rate – Momentum – Sparsity -Understa				Loss Functio	ons for Rec	onstruction	۱ -
UNIT – II	CNN	Nate – Momentum – Sparsity - Ondersta	inding Con	voiulion	5.	Periods:9			
_		Type - Convolutional Layer - Activation	Layer - Po	oling La	yer - Ful		Layer -Bat	:ch	CO2
		Common architecture and Training Patter							
UNIT – III	Optim	ization				Periods:9			<u>i</u>
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COs	ŀ	Progra	m Out	Program Specific Outcomes (PSOs)					
	PO1	PO2	PO3	PO4	PO5	P06	PSO1	PSO2	PSO3
1	2	3	2	-	-	-	1	2	1
2	3	3	3	-	1	-	2	2	1
3	2	3	2	1	2	-	2	2	1
4	3	3	3	1	3	-	2	2	1
5	3	3	3	1	3	-	2	2	1

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

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^{**}Assignment to be given from Unit-5

Department	Artificial Ir	ntelligence and Data Science	Progran	nme: M .	Tech.				
Semester	II		Course	Catego	ry: PC	*Enc	Semeste	er Exam Ty	pe: TE
0	DOG A DTGG	_	Perio	ds / We	ek	Credit	Ma	ximum Ma	rks
Course Code	P23ADT20	<i>'</i>	L	Т	Р	С	CAM	ESE	TM
Course Name	AI AND RO	DBOTIC PROCESS	3	0	0	3	40	60	100
	.4	()	AI & DS)				<u>t</u>		
Prerequisite	-								
	On comple	etion of the course, the studer	nts will b	e able t	Ю.				lapping st Level)
Course	CO1 Des	scribe RPA, where it can be applied	and how	t's imple	mented				K2
Outcomes	CO2 Des	scribe the Different Types of Variab	les, Contro	I Flow a	nd Data	Manipulation	ation Techniques K2		K2
	CO3 Ide	ntify and understand Image, Text a	nd Data Ta	ıbles Au	tomation	1.			K2
	<u> </u>	scribe how to handle the User Even					trategies.		K2
		oly techniques to deploy and mainta		, , , , , , , , , , , , , , , , , , ,					K3
UNIT – I		on to Robotic Process Automa				Periods:9			
		tion - Processes & Flowcharts - F		na Cons	structs in		Advance	d Concents	
Standardization of ousiness case - R Challenges with R	processes - R RPA Team - P PA - RPA and	PA Development methodologies - rocess Design Document/Solution emerging ecosystem.	Difference	from SI	DLC - Ro	obotic control stries best su	flow archit	ecture - RP	A CO1
JNIT – II		Introduction and Basics				Periods:9			
Flow - Control Flow The While Activity	v Activities - TI - The for Each	ser Interface - Variables - Control F ne Assign Activity - The Delay Activ Activity - The Break Activity - Data	rity - The D Manipulat	o While ion- Dat	Activity a Manip	 The If Activitual Ulation Introdu 	y-The Swi	tch Activity -	CO2
UNIT – III		 Text Manipulation - Data Manipul Automation Concepts & Tech 		nenng a	IIIU ASSE	Periods:9			
_		nd Desktop Recording - Web Recor		ıt/Outpu	t Method		raping -		
Dynamic Selectors	- Partial Select Data Tables of PDF	ted techniques - Selectors - Definin ctors - RPA Challenge - Image, Tex & PDF - Data Tables in RPA - Exce Jser Events & Assistant Bots,	kt & Advan el and Data	ced Citri Table b	x Autom asics - [ation Introduc	tion to Ima	age & Text	CO3
What are assistant element triggers - pot on a keyboard	t bots - Monito An example of event. Debug	ring system event triggers - Hotkey monitoring email - Example of mor ging and Exception Handling - Debu and Maintaining the Bot	trigger - Note in the contract of the contract	Mouse tropying each	igger - S event an itegies fo	System trigger d blocking it -	Launching	an assistar	
		Creation of Server - Using Server to					Robot from	the Server	_
Connecting a Rob backages - Deletin	ot to Server - g packages	Deploy the Robot to Server - Publi	shing and	managii	ng updat	es -Managing	packages	s - Uploadin	g CO5
LecturePeriod	s:45	TutorialPeriods:0	Practica	alPerio	ds:-0	L	ecturePe	riods:45	
Textbooks 1. Tom Taul	li, The Robotic	Process Automation Handbook: A	Guide to I	mpleme	nting RF	'A Systems, 2	020.		
		rning Robotic Process Automation"	', Packt Pu	blishing,	2018.				
Reference Book	_								
a Primer"	, Institute of Ro	Dilla, Heidi Jaynes, Lauren Livings obotic Process Automation, 1st Editionic Process Automation: Guide to	ion 2015.					nske 8 boso	mo an PD
		ntly Published, 1st Edition 2018.	Dununing S	onwale	rvobolo,	Automate Re	Pennie 10	aura & DECO	me an NE
		botic Process Automation Tools, P	Process Au	tomation	n and th	eir benefits: L	Inderstand	ling RPA an	d Intelliger
		Opportunity Holdings LLC, 1st Edi						Ŭ	J
processe	s", Packt Publi	Process Automation with Blue Prism shing, 1st Edition 2018.	n Quick Sta	rt Guide	: Create	software robo	ots and	automa	ate busines
Veb References									
•		m/rpa/robotic-process-automati	on						
https://w	www.acadamy								
•	-	.uipath.com							
3. https://w	ww.guru99.c	om/deep-learning-tutorial.html							
3. https://w 4. https://w	ww.guru99.coww.coursera.	•	ng						

5. http://neuralnetworksanddeeplearning.com/

* TE - Theory Exam, LE - Lab Exam

COs	Program Outcomes (POs)						Program Specific Outcomes (PSOs)			
	P01	PO2	PO3	PO4	PO5	P06	PSO1	PSO2	PSO3	
1	2	3	2	-	-	-	1	2	1	
2	3	3	3	-	1	-	2	2	1	
3	2	3	2	1	2	1	2	2	1	
4	3	3	3	1	3	-	2	2	1	
5	3 3		3	1	3	-	2	2	1	

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus
**Assignment to be given from Unit-5

2. 1

Department	Artificial Intelligence and Data Science	Progran	nme: M.	Tech.					
Semester	II	Course	Catego	ry Code	: PC *Er	nd Semes	ter Exam	Гуре: LE	
Course Code	P23ADP202	Perio	ds / We	ek	Credit	Credit Max		kimum Marks	
Course Code	F 23ADF 202	L	Т	Р	С	CAM	ESE	TM	
Course Name	ADVANCED DEEP LEARNING LABORATORY	0	0	4	2	50	50	100	
	(A	& DS)		·		·		·	
Prerequisite	NIL								
	On completion of the course, the stud	BT Mapping (Highest Level)							
Course	CO1 Design a simple neural network.	K6							
Outcomes	CO2 Create applications using CNN.	K6							
	CO3 Create applications using RNN and LS	K6							
	CO4 Create applications using GRU.	K6							
	CO5 Create a recommendation system.	K6							
	List of Exercises						•		

- 1. Build a simple neural network
- 2. Build a deep learning model to Classify cat and dog using CNN
- 3. Build a deep learning model to predict Stock Prices using Recurrent Neural Network
- 4. Build a deep learning model to Forecast Sales using LSTM
- 5. Build a deep learning model to predict Movie box office using GRU model
- 6. Build a deep learning model to predict Sports result Prediction using RNN and LSTM
- 7. Build a deep learning model to predict cardiovascular disease using ANN
- 8. Build a deep learning model to create an art using Style Transfer technique
- 9. Build a deep learning model to a identify traffic signs from the image
- 10. Build a deep learning model for Fashion Recommendation System

Lecture Periods:	Tutorial Periods:	Practical Periods: 30	Total Periods: 30

Reference Books

- 1. Cosma Rohilla Shalizi, "Advanced Data Analysis from an Elementary Point of View", Cambridge University Press, 2015.
- 2. Deng & Yu, "Deep Learning: Methods and Applications", Now Publishers, 2014.
- 3. Michael Nielsen, "Neural Networks and Deep Learning", Determination Press, 2015.
- 4. Josh Patterson, Adam Gibson, "Deep Learning A Practitioner's Approach", O'Reilly Media, 2017.
- 5. Nikhil Buduma, "Fundamentals of Deep Learning", O'Reilly, 2017.

Web References

- 1. https://nptel.ac.in/courses/106/106/106106212/
- 2. https://www.geeksforgeeks.org/data-analysis-visualization-python/
- 3. https://www.coursera.org/learn/python-data-analysis
- 4. https://www.python.org/
- 5. https://www.programiz.com/python-programming
 - * TE Theory Exam, LE Lab Exam

COs/POs/PSOs Mapping

COs		Progra	m Out	Program Specific Outcomes (PSOs)					
	P01	PO2	PO3	PO4	PO5	P06	PSO1	PSO2	PSO3
1	3	3	3	3	2	3	3	3	2
2	3 2		2	2	1	2	2	2	1
3	3	1	1	1	2	1	1	1	2
4	3	1	1	2	2	1	2	1	2
5	3	2	2	2	3	2	2	2	3

Correlation Level: 1 - Low, 2 - Medium, 3 - High

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	Co						
Assessment	Performan cla	ce in pract asses	ical	Model		End Semester	Total
	Conduction of practical	Record work	viva	Practical Examination	Attendance	Examination (ESE) Marks	Marks
Marks	Marks 15 5		5	15	10	50	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

Department	Artificial Intelligence and Data Science	Programme: M.Tech.						
Semester	Course Category : HS				*End Semester Exam Type: LE			
Course Code		Peri	ods / W	eek	Credit	Ma	ximum Ma	arks
Course Code	P23HSPC02	L	Т	Р	С	CAM	ESE	TM
Course Name	SEMINAR ON ICT: A HANDS-ON APPROACH	0	0	4	2	100	-	100
(Com	nmon to all M.Tech Programmes)							
Droroguicito	No Prerequisite needed							

Prerequisite	No Prerequisite needed	
	On completion of the course, the students will be able to	BT Mapping (Highest Level)
Course	CO1 Select a topic, narrowing the topic into presentation.	K2
Outcomes	CO2 State an objective and use the relevant ICT tools to make the presentation effective.	K3
	CO3 Study the topic and understanding the contributions and prepare report.	K2
	CO4 Prepare a working demo.	К3
	CO5 Prepare conclusions based on the reading of the topic and giving final Presentation.	K4

List of Experiments:

The methodology used is "learning by doing", a hands-on approach, enabling the students to follow their own pace. The teacher, after explaining the project, became a tutor, answering questions and helping students on their learning experience.

ICT skills

- Understand ICT workflow in the respective domain choosed.
- Manage multitasking.
- Deal with main issues using tech in class.
- Record, edit and deliver audio and video.
- Automate assessments and results.

Scope

- Perspective in order to design activities in class.
- Understand the process of creating audiovisuals.

Teaching tools

- Different ways to create audiovisual activities.
- Handle audiovisual editors.
- Collaborative working.
- Individualize learning experience.
- Get instant feedback from students.

Each one of the students will be assigned an ICT Topic and the student has to conduct a detailed study on the assigned topic and prepare a report, running to 30 or 40 pages for which a demo to be performed followed by a brief question and answer session. The demo will be evaluated by the internal assessment committee (comprising of the Head of the Department and two faculty members) for a total of 100 marks.

Lecture Periods: -	Tutorial Periods: -	Practical Periods: 45	Total Periods: 45	

* TE - Theory Exam, LE - Lab Exam

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COs	I	Progra	m Out	Program Specific Outcomes (PSOs)					
	P01	PO2	PO3	PO4	PO5	P06	PSO1	PSO2	PSO3
1	-	3	1	1	3	3	3	2	1
2	-	3	1	1	3	2	3	2	1
3	-	3	1	1	3	2	3	2	1
4	-	3	1	1	3	2	3	2	1
5	-	3	1	1	3	2	3	2	1

Correlation Level: 1 - Low, 2 - Medium, 3 - High

	Co	ntinuous Assess				
Assessment	Perform	nance in practical		End Semester	Total	
Accessions	Presention using ICT	Report	viva	Attendance	Examination (ESE) Marks	Marks
Marks	30	10	-	100		

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

Department	Artificial Intelligence and Data Science	Programme: M.Tech.						
Semester	Second	Course Category Code: AEC *End Semester Exam						Type: -
Course Code	P23ADC2XX	Periods / Week Cr				dit M	aximum Marks	
Course Code	FZJADGZAA	L	Т	Р	С	CAM	ES	TM
							Е	
Course Name	ABILITY ENHANCEMENT COURSE - II	0	0	4	-	100	-	100
	<u> </u>							

Students shall choose an International certification course offered by the reputed organizations like Google, Microsoft, IBM, Texas Instruments, Bentley, Autodesk, Eplan and CISCO, etc. The duration of the course is 40-50 hours specified in the curriculum, which will be offered through Centre of Excellence.

Pass /Fail will be determined on the basis of participation, attendance, performance and completion of the course. If a candidate fails, he/she has to repeat the course in the subsequent years. Pass in this course is mandatory for the award of degree.

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Department	Computer Science Engineering (Big Data Analytics)	Progran	nme: M	.Tech.				
Semester	II	Course	Catego	ory : PE	End	Semester E	xam Type	: TE
Course Code	P23BDEC02	Pe	riods / \	Week	Cre	dit	Maximum	Marks
		L	Т	Р	С	CAM	ESE	ТМ
Course Name	Web Analytics and Development	3	0	0	3	40	60	100
	(Common to M.Tech	CSE(BD	A) and	AI & DS	3)			
Prerequisite	Internet Programming							
	On completion of the course, the studer						(Highes	apping st Level)
Course	CO1 Interpret various web analytics platform				oto otroomo			(2
Outcomes	CO2 Apply various Data collection technique		and org	janize da	ata streams	j.		(3
	CO3 Interpret the benefits of qualitative analysis	-						2
	CO4 Interpret and apply various metrics to e	valuate the	perform	ance of	web data		K2,	K3
	CO5 Apply various Web analytics versions in	n real world	scenario	os			K	(3
UNIT – I	Introduction				Periods	::9	<u>.</u>	
Content charact	ess, Key terms: Site references, Keywords and Kerization terms, Conversion metrics; Categories: for web analytics, Advantages, Limitations							
}	Data Collection				Periods	s:9		
Click stream Da Brand/Advocacy	ta: Web logs, Web Beacons, JavaScript tags, Pa				i ata: Ecomn	nerce, Lead		CO2
ļ	Qualitative Analysis				Periods	s:9		
Benefits of site v data: Web logs selecting optima Link coding issu	ations: Conducting a heuristic evaluation, Beneficists; Surveys: Website surveys, post-visit survey or JavaScript's tags, Separate data serving and web analytic tool, Understanding click streamies. Web Metrics	/s, creating d data capt	and runi ure, Typ	ning a si be and s	urvey, Bene size of data	efits of surver , Innovation, definition, Us	ys. Capturin Integration	g CO3
on site, new v campaigns; Rea Introduction to h	s: Hits, Page views, Visits, Unique visitors, Uniq isits; Optimization (e-commerce, non-e- commal time report, Audience report, Traffic source report, characteristics, Need for KPI, Perspective ocomputing, HTTP (Hypertext Transfer Protocol), Web Analytics 2.0	nerce sites port, Custor of KPI, Use): Impro n campa s of KPI.	oving bo aigns, Co . Releva	ounce rates ontent repo ant Technol	s, Optimizin rt, Google ar ogies: Intern s.	g Ad Word nalytics,	s
Web analytics 1.0 Toolbar data, Par Analyzing compe	D, Limitations of web analytics 1.0, Introduction to nel data, ISP data, Search engine data, Hybrid d titive site overlap and opportunities. Google Anal fic: Organic traffic, Paid traffic; Google website of	lata, Websi lytics: Brie	te traffic introduct	analysistion and	intelligence s: Comparir working, A	analysis: CI ng long term d Words, Bo Limitations,	traffic trend enchmarking Performand	s g :e
	eriods: 45 Tutorial Periods: -	Pr	actical	Period	ls: -	Tota	l Periods:	45
2. Jure Leskovec,	nced Web Metrics with Google Analytics, Wiley Anand Rajaraman, and Jeffrey D. Ullman, "Minib Analytics 2.0, The Art of Online Accountability	ing of Mass	ive Data	asets" 2r	nd edition, C			
References Bo	oks							
 Sterne J., We Brian Clifton , " Jerri L. Ledford Pedro Sostre ,"W 	bb Metrics: Proven methods for measuring web s Advanced Web Metrics with Google Analytics", I and Joe Teixeira, "Learning Web Analytics: A E eb Analytics For Dummies", For Dummies, Sec ik, "Web Analytics 2.0: The Art of Online Accoun	Sybex, Thi Beginner's (ond Edition	rd Editio Guide to , 2012	n, 2012 Google	Analytics",	O'Reilly Me		n, 2009
Web Reference								
1. https://www.n	nygreatlearning.com/courses/big-data-analy	tics-dse						
2. https://intellip	aat.com/big-data-hadoop-training/							
3. https://www.e	edureka.co/comprehensive-hive							

COs	Prog	ram Oı	utcome	Program Specific Outcomes (PSOs)					
	PO1	PO2	PO3	PO4	PO5	PO6	PSO1		
1	3	3	2	1			-	1	3
2	3	3	2	1			-	1	3
3	3	3	2	1			-	1	3
4	3	3	2	1			-	1	3
5	3	3	2	1			-	1	3

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus *Assignment to be given from Unit-5

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Department	Artific	ial Intel	ligence and Data Science	Progran	nme: M.	Tech.					
Semester	II			Course	Catego	ry: PE	*End	d Semeste	er Exam Ty	pe: TE	
Course Code	P23A[)E205		Perio	ds / We	ek	Credit	Ма	ximum Ma	rks	
Course Code	FZJAL)L203		L	Т	Р	С	CAM	ESE	TM	
Course Name			LIZATION USING ID POWER BI	3	0	0	3	40	60	100	
Prerequisite	Pythor	n for Dat	a Science								
	On co	mpletio	n of the course, the stude	nts will b	e able t	0				Mapping est Level)	
Course	CO1 Interpret how the data can be visualized using Tableau										
Outcomes	CO2	Custon	nize and fine tune map aestheti	cs using Ta	ableau					K3	
	CO3	O3 Apply Power BI basics for generating interactive reports effectively									
	CO4	Interpre	et and apply power query and N	// language	for hand	dling tab	les		K	2, K 3	
	CO5	Apply P	ower BI concepts for data mod	eling						K3	
UNIT-I	Introd	uction i	n Tableau				Periods: 9				
shapes – Customiz	Roles – ing Map erence [Placing s – Statis Distributio	tistics using Tableau marks on a Map – Overlaying tics: Add Reference Lines Ban on-Working Reference Lines B	ds and Dis	tribution	- Adding	Reference L	n Maps – U ines -Addir	ng Referenc	:e	
UNIT-III	······································		o Power BI				Periods: 9)			
Connection of Dat	a Source	e- Repor	ting Business Intelligence (BI) esktop-Power BI Architecture-/				ced BI-Powe		cts-Power I	SI CO:	
UNIT-IV	Power	Query	And M Language		•••••	••••••	Periods: 9			······	
Query Editor, Quer IFELSE Condition	y Editor is, Trans otions Ta	User Inte form Col ableGro	ata Transformation-Shape or T rface- The Ribbon (Home, Tra umn () Types-Remove Columr oup () Table. Sort () with Type (nsform, Ad ns (), Split	d Colum Columns	ın, View s (),Repl	Tabs)-Basic lace Value()	Functions- -Table. Dis	M Language stinct Option	e- IS	
UNIT-V		/lodeling	9				Periods: 9			<u>1</u>	
Data Modeling In One, One-to-Ma	ntroduct any (or	ion -Rel Many-to	ationship, Need of Relations -One), Many-to-Many - Aut elationship Active or Inacti	oDetect tl	ne relat	ionship	, Cardinality , Create a r	in Genera new relation	onship, Ed	it CO	
Lecture Period	s: 30		Tutorial Periods: 15	Practica	al Perio	ds: -	T	otal Perio	ods: 45	L	
Text Books			· · -	1		-					

- 1. Ashutosh Nandeshwar, "Tableau Data Visualization Cookbook", Packt Publishing Ltd., 2013.
- 2. Brett Powell, Mastering Microsoft Power BI: Expert techniques for effective data analytics and business intelligence, 1st edition, 2018.
- 3. Alberto Ferrari and Marco Russo, Analyzing Data with Power BI and Power Pivot for Excel, 1st edition, 2017.

Reference Books

- 1. Scott Murray, Interactive Data Visualization for the Web: An Introduction to Designing with D3, 2nd edition, 2017.
- Cole Nussbaumer Knaflic, Storytelling with Data: A Data Visualization Guide for Business Professionals, 1st edition, 2015. 2.
- Stephen Few, Information Dashboard Design: Displaying Data for At-a-Glance Monitoring, 2nd edition, 2013.
- Nathan Yau, Data Points: Visualization That Means Something, 1st edition, 2013.
- 5. Edward R. Tufte, The Visual Display of Quantitative Information, 2nd edition, 2001

Web References

- https://www.tableau.com/
- https://www.guru99.com/what-is-tableau.html
- https://www.datacamp.com/tutorial/data-visualisation-powerbi
- https://learn.microsoft.com/en-us/power-query/power-query-ui
- https://www.tutorialspoint.com/power_bi/power_bi_data_modeling.html

COs	ŀ	Progra	m Out	comes	(POs)	Program Specific Outcomes (PSOs)			
	PO1	PO2	PO3	PO4	PO5	P06	PSO1	PSO2	PSO3	
1	2	3	2	-	-	-	1	2	1	
2	3	3	3	-	1	-	2	2	1	
3	2	3	2	1	2	-	2	2	1	
4	3	3	3	1	3	-	2	2	1	
5	3	3	3	1	3	-	2	2	1	

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

5. V) / -

^{**}Assignment to be given from Unit-5

Department	Artifi	cial Intelligence and Data Science	Progran	nme: M.	Tech				
Semester	II		Course	Catego	ry: PE	*End	l Semeste	r Exam Ty	⁄ре: ТЕ
Course Code	D23A	DE206	Perio	ods / We	ek	Credit	Ma	ximum Ma	rks
Course Code	1 234		L	Т	Р	С	CAM	ESE	TM
Course Name	PRE	DICTIVE MODELLING	3	0	0	3	40	60	100
		(/	AI & DS)		<u></u>		<u>-</u>		.
Prerequisite	NIL								
		lapping est Level)							
Course	CO1	Interpret the steps for predictive modelin	g and real	ize it in r	eal world	d			K2
Outcomes	CO2	Interpret various terminologies and steps	s for predic	tive mod	leling				K2
	CO3	Apply and improvise machine learning to	echniques	for predic	ctive mo	deling.			K3
	CO4	Apply Predictive Modeling Markup Lang	uage for m	aking pro	edictive i	modeling easi	ier		K3
	CO5	Apply predictive modeling techniques in	real world	case stu	dies				K3
UNIT-I	Intro	duction				Periods: 9			
or predictive anal	ytics.	-processing - Data cleaning - Data Partiti	oning - Bu	ilding a r	nodel - S			stical mode	ls CO1
UNIT-II		ictive Modeling				Periods: 9			
		Over fitting –Oversampling –Multiple Reg s - Model specification - Model selection				works (MLP)	- Variable i	mportance-	CO2
UNIT-III	Predi	ictive Models				Periods: 9			
/lodel – Regressi	on Mode Scoreca	ng Models –Decision Trees- Ruleset Mod Is – Regression Trees – Classification & F ards – Support Vector Machines – Time S odel.	Regressior	n Trees (ČART) -	- Logistic Řeg	ression - N	/lultiple	CO3
UNIT-IV	Predi	ctive Modeling Markup Language				Periods: 9			
ntroduction to PM Model Support – N		MML Converter - PMML Structure – Data erification.	Manipulati	on in PM	IML – PN	MML Modeling	g Techniqu	es - Multiple	CO4
UNIT-V	Tech	nologies And Case Studies				Periods: 9			i
	ner – IBN	M SPSS Statistics- IBM SPSS Modeler – e study with modeling and analysis.	SAS Entei	prise Mi	ner – Ap	ache Mahout	– R Progra	ımming	COS
Lecture Period	-l 4F	Tutorial Periods: 15	Practic	- I D: -			otal Perio		

Lecture Periods: 45 Tutorial Periods: 15 Practical Periods: - Total Periods: 60

Text Books

- 1. Statistical and Machine-Learning Data Mining Techniques for Better Predictive Modeling and Analysis of Big Data, Second Edition, 2011
- 2. Predictive Modeling with SAS Enterprise Miner Practical Solutions for Business Applications, Third Edition, 2017
- 3 Mastering Predictive Analytics with R.Second Edition: Machine learning techniques for advanced models, 2017

Reference Books

- 1. Kattamuri S. Sarma, "Predictive Modeling with SAS Enterprise Miner: Practical Solutions for Business Applications", 3rd Edition, SAS Publishing, 2017.
- 2. Alex Guazzelli, Wen-Ching Lin, Tridivesh Jena, James Taylor, "PMML in Action Unleashing the Power of Open Standards for Data Mining and Predictive Analytics", 2nd Edition, Create Space Independent Publishing Platform,2012
- 3. Ian H. Witten, Elbe Frank , "Data Mining: Practical Machine Learning Tools and Techniques", Morgan Kaufmann Series in Data Management Systems, Morgan Kaufmann, 3rd Edition, 2011
- 4. Eric Siegel, "Predictive Analytics: The Power to Predict Who Will Click, Buy, Lie, or Die", 2nd Edition, Wiley, 2016.
- 5. Conrad Carlberg, "Predictive Analytics: Microsoft Excel", 1st Edition, Que Publishing, 2012.

Web References

- 1.https://nptel.ac.in/courses/108108111/
- 2.https://www.coursera.org/learn/predictive-modeling-analytics
- 3.https://bookdown.org/egarpor/PM-UC3M/
- 4.https://cics.nd.edu/research/applications/materia
- 5.https://www.netsuite.com/portal/resource/articles/financial-management/predictive-modeling.shtml

2. 1

COs	ŀ	Progra	m Out	comes	(POs)	Program Specific Outcomes (PSOs)			
	P01	PO2	PO3	PO4 PO5 PO6		PSO1	PSO2	PSO3		
1	2	3	2	-	-	-	1	2	1	
2	3	3	3	-	1	-	2	2	1	
3	2	3	2	1	2	-	2	2	1	
4	3	3	3	1	3	-	2	2	1	
5	3	3	3	1	3	-	2	2	1	

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Assessment		Continu	ous Asse	(CAM)	End Semester	Total	
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

Department	Artific	cial Intel	ligence and Data Science	Progran	nme: M.	Tech				
Semester	II			Course	Catego	ry:PE	*End	d Semeste	r Exam Typ	e: TE
Course Code	РЭЗА	DE207		Perio	ds / We	eek	Credit	Max	imum Mark	(S
Course Code	1 23A	DLZUI		L	Т	Р	С	CAM	ESE	TM
Course Name	NEXT SYST		ATION DATABASE	3	0	0	3	40	60	100
	•			AI & DS)						
Prerequisite	NIL									
		•	n of the course, the stude						BT Ma (Highes	
Course Course Interpret market and technology leading to today's next generation databases.									K	2
Outcomes	CO2	Apply Ha	doop architecture with querying	g in various	Hadoop	compo	nents		K	3
	CO3	Identify X	ML and JSON Document Data	bases					K	2
	CO4	Identify d	atabase applications oriented t	o Graph ar	nd Colum	nn datab	ases		K	2
	CO5	Apply the Cassand	Distributed Database patterns	and consi	stency m	nodels in	MongoDB, H	Base and	K	3
UNIT-I	Datak	ase Rev	olution				Periods: 9			
Database Wars - C	lient-se	rver Comp	Database Revolution: Relation outing – Object Oriented Programent Databases – NEWSQL.							CO1
UNIT-II	Hado	ор: Оре	n-Source Google Stack				Periods: 9			
			– Hadoop's Architecture – Wo educe – Hbase – Pig - Hive: Qu							CO2
UNIT-III	Docu	ment Da	itabases				Periods: 9		<u>i</u>	
			ındards – XML support in Relat ses – Early JSON Databases –					ases – JSO	N and AJAX	CO3
UNIT-IV	Grapl	h and Co	olumn Databases				Periods: 9			
Internals – Graph C	Compute	Engines.	or Graphs – RDF and SPARC Column Databases: Data Ware Q, C-Store and Vertica – Colun	ehouse Sch	iema – C	columna				CO4
UNIT-V			atabase Patterns and Con				Periods: 9			
	e – Cas		buted Relational Databases – onsistency Models: Types of C							CO5
Lecture Period	s: 45		Tutorial Periods: 15	Practic	al Perio	ds: -	T	otal Perio	ds: 60	
Text Books							······································			

- 1. Enhancing Availability for NoSQL Database Systems using Failover Techniques, Priyanka Gotter, Kiranbir Kaur, Tanveer Kaur
- 2. NoSQL A complete guide, 2021 edition

Reference Books

- 1. Guy Harrison, "Next Generation Databases: NoSQL, NewSQL, and Big Data", Apress Publisher, 2016
- 2. Chanchal Singh and Manish Kumar, "Mastering Hadoop 3: Big data processing at scale to unlock unique business insights", Packt Publishing, 2019.
- 3. Subhashini Chellappan, Dharanitharan Ganesan, "MongoDB Recipes: With Data Modeling and Query Building Strategies", Apress Publisher, 2019
- 4. Jeff Friesen, "Java XML and JSON: Document Processing for Java SE", Apress Publisher, 2019

Web References

- https://www.researchgate.net/publication/221214756 The Next Database Revolution
- https://cloudxlab.com/blog/big-data-solution-apache-hadoop-and-spark/
- https://www.mongodb.com/document-databases
- https://www.geeksforgeeks.org/document-databases-in-nosql/
- https://towardsdatascience.com/cap-theorem-and-distributed-database-management-systems-5c2be977950e

TE - Theory Exam, LE - Lab Exam



COs	1	Progra	m Out	comes	(POs)	Program Specific Outcomes (PSOs)			
	PO1	PO2	PO3	PO4	PO5	P06	PSO1	PSO2	PSO3	
1	2	3	2	-	-	-	1	2	1	
2	3	3	3	-	1	-	2	2	1	
3	2	3	2	1	2	•	2	2	1	
4	3	3	3	1	3	-	2	2	1	
5	3	3	3	1	3	-	2	2	1	

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Evaluation Method

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus **Assignment to be given from Unit-5

5. X } \ _

Department	Artificial Inte	Iligence and Data Science	Progran	nme: M.	Tech				
Semester	II		Course	Catego	ry: PE	*End	d Semeste	r Exam Ty	pe: TE
Course Code	P23ADE208		Perio	ds / We	ek	Credit	Ma	ximum Mar	ks
	0,		L	Т	Р	С	CAM	ESE	TM
Course Name	ADVANCED A	ALGORITHMS	3	0	0	3	40	60	100
		(/	AI & DS)						
Prerequisite	-								
	-	on of the course, the studer							apping st Level
Course	CO1 Interpret	algorithm evaluation techniques	and ensu	re algori	thm corr	ectness		l	√2
Outcomes	CO2 Apply gr	eedy algorithms to solve probler	ns					I	₹3
	CO3 Apply div	vide and conquer algorithm to so	olve real w	ord prob	lems			ı	∢ 3
	CO4 Apply ne	twork flow algorithms to handle	graph rela	ted real	world pro	blems		ı	√ 3
	CO5 Interpret	NP-complete problems and app	ly them in	real wor	ld scena	rios		K2	., K 3
UNIT-I	Periods: 9								
ueues and Stac reedy Algorith rgument – The I	efinitions and Appli ks – Testing Bipar ms: Interval Sched Minimum Spanning	Greedy Algorithms cations – Graph connectivity and titeness: An application of Bread fuling: The Greedy Algorithm State Tree Problem – Implementing I	Ith First se ays Ahead	arch. – Optim	al Cachiı	ng: A More C	Traversal	change	CC
	and Data Compre					Periods:			
UNIT-III	Divide and C	•	o Dolotion		ation love		_	acat Dair of	: [
oints – Integer N Dynamic Program	Multiplication Dyna nming: Memoizatio acks: Adding a var	rt Algorithm – Further Recurrence mic Programming: Weighted Ir n or Iteration over Subproblems iable – Shortest Paths in a Grap	nterval Sch – Segmer	neduling: nted Lea	A Recui st Squa	rsive Procedu res : Multi-wa	ure – Princi _l ny Choices	oles of - Subset	CO
UNIT-IV	Network Flov	V				Periods:	9		L
		e Ford-Fulkerson Algorithm – M pplication: The Bipartite Matchin							CO
UNIT-V	NP and Com		Periods:	9		<u>i</u>			
		cient Certification and the Defini oring – Co-NP and the Asymme		– NP-Co	omplete	Problems – S	Sequencing	Problems –	CO
Lecture Peri	ods: 15	Tutorial David do: 45	Practic	al Dar	lada.	Т	otal Peri	odei 60	
Lecture Peri	0us. 43	Tutorial Periods: 15	Fractic	ai ren	10as: -	' I	otal Peri	ous. ou	
ext Books	ous. 43	Tutorial Periods: 15	Fractic	ai Per	ious: -		otal Peri	ous. 60	

- 2. Algorithms, Robert Sedgewick and Kavin Wayne, 4th edition

Reference Books

- 1. Jon Kleinberg, ÉvaTardos, "Algorithm Design",Pearson Education Limited 2014.
- 2.Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, Clifford Stein, "Introduction to Algorithms", MIT Press, 2009.
 3. Ellis Horowitz, SartajSahni and SanguthevarRajasekaran, "Fundamentals of Computer Algorithms", Second Edition, Universities Press, Hyderabad, 2008.
- 4. AnanyLevitin, "Introduction to the Design and Analysis of Algorithms", Third Edition, Pearson Education Asia, 2008.

Web References

- https://www.scaler.com/topics/analysis-of-algorithm/
- https://www.codingninjas.com/studio/library/greedy-algorithm-in-graph-theory
- https://www.tutorialspoint.com/data_structures_algorithms/divide_and_conquer.htm
- https://www.cs.cmu.edu/~avrim/451f11/lectures/lect1025.pdf
- https://cseweb.ucsd.edu//classes/sp05/cse101/Day19NP.pdf

COs		Progra	m Out	comes	(POs)		ram Spe omes (P	
	PO1	PO1 PO2 PO3 PO4 PO5 PO6						PSO2	PSO3
1	2	3	2	-	-	-	1	2	1
2	3	3	3	-	1	-	2	2	1
3	2	3	2	1	2	-	2	2	1
4	3	3	3	1	3	1	2	2	1
5	3	3	3	1	3	-	2	2	1

Correlation Level: 1 - Low, 2 - Medium, 3 - High

Assessment		Continu	ous Asse	essment Marks	(CAM)	End Semester	Total
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance	Examination (ESE) Marks	Marks
Marks	1	0	15	10	5	60	100

^{*} Application oriented / Problem solving / Design / Analytical in content beyond the syllabus **Assignment to be given from Unit-5

Department	Artifici	ial Intelli	gence and Data Science	Progran	gramme: M.Tech.					
Semester	1/11			Course	Catego	ory : AE	C *Er	id Semestei	Exam Ty _l	oe: -
Course Code	РЭЗЛ	CTX01		Perio	ods / W	eek	Credit	: Ma	ximum Ma	rks
				L	Т	Р	С	CAM	ESE	TM
Course Name	ENGLI WRITII		RESEARCH PAPER	2	0	0	-	100	-	100
			(Commo	on to all M	.Tech F	⊃rogran	nme)			
Prerequisite	No P	rerequisi	te needed							
	On co	ompletio	n of the course, the stude	nts will b	e able	to			BT Ma (Highest	
_	CO1	Understa	nd that how to improve your w	riting skills	and leve	el of read	dability.		K2	
Course Outcomes	CO2	Learn abo	out what to write in each section	n.					K1	
Outcomes	CO3	Understa	nd the skills needed when writ	ng a Title.					K2	
	CO4	Understa	nd the skills needed when writ	ng the Con	clusion.	•			K2	
	CO5	Ensure th	e good quality of paper at very	/ first-time s	submiss	ion.			K3	
UNIT- I	Introdu	uction to	Research Paper Writing				Periods	: 6	·····	
Removing Redun	dancy, A	Avoiding A	der, Breaking up long sentence mbiguity and Vagueness.	es, Structur	ing Para	agraphs	•		oncise and	CO1
UNIT- II	1	entation					Periods	_	, 5	
Abstracts, Introdu		Highlightii	ng Your Findings, Hedging and	Criticizing	, Parapr	nrasing a	and Plagian	sm, Sections	or a Paper,	CO2
UNIT- III	Title \	Writing S	Bkills				Periods	: 6		
	ded whe	en writing	a Title, key skills are needed w ting a Review of the Literature							CO3
UNIT- IV	Resu	lt Writing	g Skills				Periods	: 6		
			Methods, skills needed when w writing the Conclusions.	riting the R	Results,	skills are	needed wh	nen writing th	е	CO4
UNIT- V		cation S	-				Periods	_		
Jseful phrases, c	hecking	Plagiarisn	n, how to ensure paper is as g	··•			the first- tim			CO5
Lecture Perio	ds: 30		Tutorial Periods: -	Practic	al Perio	ods: -		Total Perio	ods: 30	
Reference Boo										
2. Day R, "How to 3. Goldbort R, "W	Write a	nd Publish Science",	ting Research Papers", Spring n a Scientific Paper", Cambridg Yale University Press (Availal g for the Mathematical Science	ge Universit	y Press gle Bool	, 2006. ks), 2006	3.	ondon, 2011.		

Assessment	Cor	Continuous Assessment Marks (CAM)							
	Assignment 1	Assignment 2	Test 1	Test 2	Attendance				
Marks	20	20	25	25	10	-	100		



Department	Artific	ial Intellig	ence and Data Science	Progran	nme: N	/I.Tech	١.				
Semester	1/11			Course	Categ	ory : A	EC	*End	Semester	Exam Ty	pe: -
Course Code	D23/	ACTX02		Perio	ds / W	/eek	Cr	edit	Ma	aximum Marks	
Course Code	ГДЭР	ACT AUZ		L	Т	Р		С	CAM	ESE	TM
Course Name	DISA	STER MA	NAGEMENT	0	0	2		-	100	-	100
	.4		(Commo	n to all M	.Tech	Progra	ımme)				
Prerequisite	No P	rerequisite	needed								
	On c	ompletion	of the course, the stude	nts will b	e able	to				BT Ma (Highes	
_	CO1	Ability to s	ummarize basics of disaster.							K	1
Course Outcomes	CO2	humanitar	xplain a critical understanding ian response.							K	2
	CO3	from multi	lustrate disaster risk reductior ple perspectives.			•		•		К3	
	CO4	relevance	lescribe an understanding of s in specific types of disasters a	and conflict	situatio	ons.			•	K3 K3	
	CO5										
UNIT- I	<u> </u>	duction					Perio				
Disaster: Definitio Difference, Natur			nificance; Difference between itude.	Hazard An	d Disas	ster; Na	itural and	Mann	nade Disas	ters:	CO.
UNIT- II			s of Disasters and Hazard	ds			Perio	ds: 6	3		
Cyclones, Tsunar	ge, Loss nis, Floo	s of Humar ods, Drough	n and Animal Life, Destruction ats and Famines, Landslides a spills, Outbreaks Of Disease A	n of Ecosy nd Avalanc	hes, Ma	an-mac	Disasters le disaste	s: Ear	thquakes, \		
UNIT- III	7	•••••	Areas in India	p	,		Perio	ds: 6	3		i
			e To Floods and Droughts, La Tsunami; Post-Disaster Disea				; Areas Pr	one T	o Cyclonic	and Coasta	co3
UNIT- IV	Disa	ster Prepa	aredness and Manageme	nt			Perio	ds: 6	3		
			mena Triggering a Disaster o Agencies, Media Reports: Go							te Sensinç), CO 4
UNIT- V	Risk	Assessm	ent				Perio	ds: 6	3		<u>i</u>
			s, Disaster Risk Reduction, G n Risk Assessment and Warr								
Lecture Perio	ds: 30		Tutorial Periods: -	Practic	al Peri	iods:	•	٦	Total Perio	ods: 30	
Reference Boo	ks	<u>t</u>						<u>L</u>			
2. NishithaRai, Si	ngh AK	, "Disaster I	n And Management Text And Management in India: Perspec Mitigation Experiences And Re	tives, issue	es and	strateg	es", New	Royal	book Com		

	Conti	nuous Assess	CAM)	End			
Assessment	Assignment 1	Assignment 2	Test 1	Test 2	Attendance	Semester Examination (ESE) Marks	Total Marks
Marks	20	20	25	25	10	-	100

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Department	Artific	ial Intelligence and Data Science	Prograr	nme: M .	Tech.					
Semester	1/11		Course	Catego	ry : AE (C *En	d Semeste	r Exam Ty	pe: -	
Course Code	P23A	CTX03	Perio	ods / We	eek	Credit	Ma	aximum Ma	arks	
Course Code	1 237	101703	L	Т	Р	С	CAM	ESE	TM	
Course Name	SANS	KRIT FOR TECHNICAL	0	0	2	-	100	-	100	
	KNOW	/LEDGE								
		(Comm	on to all M	I.Tech F	rogram	ıme)				
Prerequisite	No P	rerequisite needed								
	On c	ompletion of the course, the stude		e able t	to			BT Ma (Highes		
	CO1	Understanding basic Sanskrit language) .					K	2	
Course Outcomes	CO2	Write sentences						K2		
Outcomes	CO3	Know the order and roots of Sanskrit.						К3		
	CO4	Know about technical information abou	t Sanskrit li	terature				K	3	
	CO5	CO5 Understand the technical concepts of Engineering.								
UNIT- I	Alph	abets				Periods:	6	<u> </u>		
Alphabets in San	skrit.				<u></u>				CO1	
UNIT- II	Tens	es and Sentences				Periods:	6			
Past/Present/Fut	ure Tens	se - Simple Sentences.							CO2	
UNIT- III	Orde	r and Roots				Periods:	6			
_	.	ots of Engineering-Electrical, Mechanica	I, Architectu	ıre, Math	ematics					
		,	,	,					CO	
UNIT- IV	<u>i</u>	krit Literature				Periods:	6			
Technical information	ation abo	out Sanskrit Literature.							CO	
UNIT- V	Tech	nical Concepts of Engineering				Periods:	6		<u>i</u>	
echnical concep	ots	-							CO	
	yde: 30	Tutorial Periods: -	Practic	al Pario	vye		Total Peri	ods: 30	CO	
Lecture Perio		i atoriar i crioas.	i iuotio	u 0110	, u.u.		. J.u C.			

	Conti	nuous Assess	ment N	larks (CAM)	End	
Assessment	Assignment 1	Assignment 2	Test 1	Test 2	Attendance	Semester Examination (ESE) Marks	Total Marks
Marks	20	20	25	25	10	•	100

5. X S .__

Department	Artific	ial Intelli	gence and Data Science	Programme: M.Tech.									
Semester	1/11			Course	Catego	ry : AE	C *End S	Semester Ex	kam Type:	-			
Course Code	P23A	CTX04		Perio	ds / We	eek	Credit	Credit Maximum		Marks			
Course Coue	1 237	101X0+		L	Т	Р	С	CAM	ESE	TM			
Course Name	VAL	JE EDUC	ATION	0 0 2 - 100 -									
			(Commo	n to all M	.Tech F	rogran	nme)	<u>i</u>					
Prerequisite	No P	rerequisit	e needed										
	On c	ompletio	n of the course, the stude	nts will b	e able	to			BT Ma (Highest				
_	CO1	Knowledg	ge of self-development.						K	2			
Course Outcomes	CO2	Learn the	importance of Human values.						K1				
Outcomes	CO3	Developir	ng the overall personality.						K:	3			
	CO4	Developir	ng Character and Competence						K	3			
UNIT- I	Value	es and Se	elf Development				Periods	: 6					
	Standard	ds and prin	ll values and individual attitude ciples. Value judgments of Wo y.							CO1			
UNIT- II	Culti	vation of	Values				Periods	: 6					
			Sense of duty. Devotion, Self-r National Unity. Patriotism. Love				entration. T	ruthfulness, (Cleanliness.	CO2			
UNIT- III	Perso	onality D	evelopment				Periods	: 6					
and Kindness. Av	oid fault	Thinking.	ent-Soul and Scientific attitude Free from anger, Dignity of lab love for truth. Aware of self-des	our. Unive	rsal brot	her hoo	d and religion	ous tolerance	. True	CO3			
UNIT- IV	Char	acter Dev	/elopment				Periods	: 6					
Character and Co Nonviolence, Hur			ooks vs Blind faith. Self-manag	ement and	l Good h	nealth. S	cience of re	eincarnation.	Equality,	CO4			
Lecture Perio	ds: 30		Tutorial Periods: -	Practica	al Perio	ods: -		Total Peri	ods: 30				
Reference Boo	ks						·						
	1/ "\/_1.	use and Et	hics for organizations Theory a	and practice	o" Ovfo	rd I Inive	reity Prace	Now Dolhi					

	Conti	nuous Assess	ment N	larks (CAM)	End	
Assessment	Assignment 1	Assignment 2	Test 1	Test 2	Attendance	Semester Examination (ESE) Marks	Total Marks
Marks	20	20	25	25	10	-	100

5. V ?/ ·

Department	Artific	ial Intelligence and Data Science	Progran				7			
Semester	1/11		Course			ΕĊ		nester Exa		
Course Code	P23A	CTX05	Perio	ds / W	eek/		Credit		ximum Ma	arks
			L	Т	Р		С	CAM	ESE	TM
Course Name	CON	STITUTION OF INDIA	0	0	2		-	100	-	100
		(Commo	n to all M	.Tech	Progra	amm	ne)			4
Prerequisite	No Pr	erequisite needed								
		ompletion of the course, the stude							BT Ma (Highes	
Course		Discuss the growth of the demand for civarrival of Gandhi in Indian politics.	Ū					efore the	K	3
Outcomes		Discuss the intellectual origins of the fra conceptualization of social reforms leadi	ng to revol	ution in	India.				К3	
		Discuss the circumstances surrounding [CSP] under the leadership of Jawaharla direct elections.	al Nehru ar	nd the e					K	3
	CO4 Discuss the passage of the Hindu Code Bill of 1956.									3
CO5 Discuss the administration and Election commission									K	3
UNIT- I		y of Making of The Indian Constitu	tion			l	Periods: 6	5		
	· v · · · · · · · · · · · · · · · · · · ·	ee, (Composition & Working).				······································				CO1
UNIT- II	:	sophy of The Indian Constitution				ا	Periods: 6	3		
Preamble, Salien	· · · · · · · · · · · · · · · · · · ·									CO2
UNIT- III		ours of Constitutional Rights and I					Periods: 6			
		nt to Equality, Right to Freedom, Right ag to Constitutional Remedies, Directive Pri							ltural and	CO3
UNIT- IV	Orga	ns of Governance				ı	Periods: 6	5		•
		Qualifications and Disqualifications, Pow intment and Transfer of Judges, Qualifications						, Governor,	Council o	f CO4
UNIT- V	Loca	Administration and Election Com	mission			ı	Periods: 6	5		
Municipal Corpora and role. Block lomportance of gra	ation. Pa evel: Organs ass root	ead: Role and Importance, Municipalities chayati raj: Introduction, PRI: Zila Pachay ganizational Hierarchy (Different departr democracy. Election Commission: Role and Bodies for the welfare of SC/ST/OB	at. Elected ments), Vill and Fund	l official lage lev ctioning	s and t vel: Ro	heir de o	roles, CEO f Elected a ection Comr	Zila Pachay nd Appoint nissioner a	vat: Positio ed officials nd Electio	n CO 5 s,
Lecture Perio	ds: 30	Tutorial Periods: -	Practica	al Peri	ods:	-	1	otal Perio	ods: 30	
Reference Boo										
2. Dr.S.N.Busi, D 3. M.P. Jain, India	r.B. R.Ar an Const	lia, 1950(Bare Act), Government Publicat mbedkar framing of Indian Constitution, 1 titution Law, 7 th Edition, Lexis Nexis, 2014 to the Constitution of India, Lexis Nexis,	st Edition, 4.	2015.						

Assessm ent	Con	tinuous Asses	End Semester Examination (ESE) Marks	Total Marks			
ent	Assignment 1	Assignmen t 2	Test 1	Test 2	Attendance		
Marks	20	20	25	25	10	-	100



	Artific	al Intelligence and Data Scie	nce Progran	mme: M	l.Tech.						
Semester	1/11		Course	Catego	ory : AE	C *End Se	mester Exa	ım Type:	-		
-Course Code	БЭЗΛ	CTX06	Perio	ods / W	eek	Credit	Max	ximum M	arks		
-Course Code	FZJA	C1X00	L	Т	Р	С	CAM	ESE	TM		
Course Name	PED/	GOGY STUDIES	0	0 0 2 - 100 -							
	<u> </u>	(Co	ommon to all M	1.Tech I	Progran	nme)					
Prerequisite	No Pr	erequisite needed									
	On co	ompletion of the course, the s	students will b	e able	to			BT Ma (Highes			
Course		What pedagogical practices are be developing countries?							(2		
Outcomes		What is the evidence on the effect conditions, and with what populati	ion of learners?					K	(2		
		How can teacher education (curriculum and practicum) and the school curriculum and guidance materials best support effective pedagogy?									
	CO4	Illustrate Professional developmer	nt					K	K3		
	CO5	Identify Research gaps and Future	e Directions					K	(3		
UNIT- I	Introd	uction and Methodology				Periods:	6				
Nime and rational	- Dalie						—				
		r background, Conceptual framewo amework, Research questions – O					rriculum, Tea	acher	СО		
education - Conce UNIT- II	eptual fra Them	amework, Research questions – O atic Overview	verview of metho	odology	and Sea	rching. Periods:	6		СО		
education - Conce UNIT- II Pedagogical pract	Them	amework, Research questions – O	verview of metho	odology	and Sea	rching. Periods:	6				
education - Conce UNIT- II Pedagogical pract	Them tices are	amework, Research questions – O atic Overview	verview of metho	ssrooms	and Sea	rching. Periods:	6 ies - Curricu				
education - Conce UNIT- II Pedagogical pract Feacher education UNIT- III Methodology for the practicum) and the and nature of the	Them tices are n. Evident he in defers to so to	amework, Research questions – Onatic Overview being used by teachers in formal	and informal class Pedagogical Foodback studies - s best support ef	essrooms Practice How ca	in develes es n teache pedagog	Periods: Periods: Periods: r education y? - Theory of	6 ies - Curricu 6 curriculum a of change - S	lum,	CO2		
education - Conce UNIT- II Pedagogical pract Feacher education UNIT- III Methodology for the practicum) and the land nature of the	Them tices are n. Evide he in de e school body of es and b	amework, Research questions – Onatic Overview being used by teachers in formal ence on The Effectiveness of oth stage: quality assessment of in curriculum and guidance material evidence for effective pedagogical	and informal class Pedagogical Foodback studies - s best support ef	essrooms Practice How ca	in develes es n teache pedagog	Periods: Periods: Periods: r education y? - Theory of	6 ies - Curricu 6 curriculum a of change - S al approache	lum,	CO2		
education - Conce UNIT- II Pedagogical practical Feacher education UNIT- III Methodology for the practicum) and the practicum and the practicum of the reachers' attitude	Them tices are n. Evide he in de e school body of es and bo Profe	amework, Research questions – Onatic Overview being used by teachers in formal ence on The Effectiveness of oth stage: quality assessment of in curriculum and guidance material evidence for effective pedagogical eliefs and Pedagogic strategies	and informal class Pedagogical Findled studies - s best support efin practices - Pedagogical Findled studies - s best support efin practices - Pedagogical Findled studies - Pedagogical Findled studies - Pedagogical Findled studies - Pedagogical Findled studies - Pedagogical Findled studies - Pedagogical Findled studies - Pedagogical Findled studies - Pedagogical Findled studies - Pedagogical Findled studies - Pedagogical Findled studies - Pedagogical Findled studies - Pedagogical Findled studies - Pedagogical Findled studies - S best support efindled studies - Pedagogical Findled studies - S best support efindled studies - Pedagogical Findled studies - S best support efindled studies - Pedagogical Findled studies - S best support efindled studies - Pedagogical Findled studies - S best support efindled studies - Pedagogical Findled st	Practice How ca ifective pagogic th	in develues es n teacher pedagographeory and	Periods: reducation by? - Theory of pedagogical Periods: support - Si	6 ies - Curricu 6 curriculum a of change - S al approache 6 upport from tl	nd Strength s -	CO		
education - Conce UNIT- II Pedagogical practical Feacher education UNIT- III Methodology for the practicum) and the practicum and the practicum of the reachers' attitude	Them tices are n. Evide he in de e school body of es and be Profe elopmen ommuni	amework, Research questions – Onatic Overview being used by teachers in formal ence on The Effectiveness of poth stage: quality assessment of in curriculum and guidance material evidence for effective pedagogical eliefs and Pedagogic strategies ssional Development alignment with classroom practic by - Curriculum and assessment - E	Pedagogical Find Included studies - s best support of practices - Pedagogical Find Included studies - s best support of practices - Pedagogical Find Included studies - Pedagogical Find Included Studies - Pedagogical Find Included Studies - Pedagogical Find Included Fi	Practice How ca ifective pagogic th	in develues es n teacher pedagographeory and	Periods: reducation by? - Theory of pedagogical Periods: support - Si	6 curriculum a of change - Sal approache ppport from the class sizes	nd Strength s -	CO		
education - Conce UNIT- II Pedagogical pract Feacher education UNIT- III Methodology for the practicum) and the practicum and the practicum and researchers' attitude UNIT- IV Professional development and the concept and th	Them tices are n. Evide he in de e school body of es and be Profe elopmentommuni Rese	amework, Research questions – Onatic Overview being used by teachers in formal ence on The Effectiveness of oth stage: quality assessment of in curriculum and guidance materials evidence for effective pedagogical eliefs and Pedagogic strategies ssional Development alignment with classroom practic	Pedagogical Facilided studies - s best support of practices - Pedagogical Facilided studies - s best support of practices - Pedagogical Facilided studies - s best support of practices - Pedagogical Facility of the studies of the st	Practice How ca ifective pagogic the	in develues in teacher bedagog neory and rt – Peer ed resou	Periods: reducation by - Theory of pedagogical pedagogical resupport - Sirces and large	6 ies - Curricu 6 curriculum a of change - S al approache 6 upport from tl e class sizes	lum, nd Strength s -	CO2		

- 1. Ackers J, Hardman, F, "Classroom interaction in Kenyan primary schools, Compare", 31(2): 245-261, 2001.
- 2. Agrawal M, "Curricular reform in schools: The importance of evaluation, Journal of Curriculum Studies", 36(3):361-379, 2004.
- 3. Akyeampong K, "Teacher training in Ghana-does it count? Multi-site teacher education research project (MUSTER) country report", London, DFID, 2003.
- 4. Akyeampong K, Lussier K, Pryor J, Westbrook J, "Improving teaching and learning of basic maths and reading in Africa: Does teacher preparation count?", International Journal Educational Development, 33(3): 272–282, 2013.
- 5. Alexander RJ, "Culture and pedagogy: International comparisons in primary education", Oxford and Boston: Blackwell, 2001.
- 6. Chavan M, "Read India: Amass scale, rapid, 'learning to read' campaign", 2003.
- 7. www.pratham.org/images/resource%20working%20paper%202.pdf.

	Conti	nuous Assess	CAM)	End			
Assessment	Assignment 1	Assignment 2	Test 1	Test 2	Attendance	Semester Examination (ESE) Marks	Total Marks
Marks	20	20	25	25	10	-	100

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Department	Artific	ial Intelli	gence and Data Science	Programme: M.Tech.								
Semester	1/11			Course	Catego	ry : AE (C *End Se	mester Exa	am Type: ·	•		
Course Code	P23A	CTX07		Perio	ds / We	eek	Credit		ximum Ma	arks		
				L	Т	Р	С	CAM	ESE	TM		
Course Name	STRE	ESS MAN	IAGEMENT BY YOGA	0	100	-	100					
	à		(Comm	on to all M	.Tech F	rogram	ime)					
Prerequisite	No P	rerequisit	e needed									
	On c	ompletio	n of the course, the stude	ents will b	e able 1	to			BT Mappir (Highest Le			
_	CO1	Develop	healthy mind in a healthy body	thus impro	ving soc	ial healt	h also		K2			
Course Outcomes	CO2	Improve (efficiency.						K2			
Outcomes	CO3	CO3 Understand Asan and Pranayam								2		
	CO4	CO4 Apply Asanas										
	CO5	CO5 Apply Pranayam										
UNIT- I		duction					Periods:	6				
Definitions of Eig	ht parts	of yoga. (A	Ashtanga).							CO1		
UNIT- II	Do`s	and Don	't's in Life				Periods:	6				
Yam and Niyam bramhacharya			s in life - i) Ahinsa, satya, astho	eya, bramh	acharya	and apa	rigraha, ii) A	hinsa, satya	ı, astheya,	CO2		
UNIT- III	Asan	and Pra	nayam				Periods:	6		<u>i</u>		
Asan and Pranageffects-Types of			a poses and their benefits for	mind & bo	dy - Reg	gularizati	on of breatl	ning techniq	ues and its	CO3		
UNIT- IV	Asan	Practice	es				Periods:	6				
Practice on Vario	us yoga	poses								CO4		
UNIT- V	Pran	ayam Pra	actices				Periods:	6		.i		
Practice on vario	us prana	iyam								COS		
	v4e- 30		Tutorial Periods: -	Practic	al Perio	ods: -		Total Perio	ods: 30			
Lecture Perio	us. Ju						<u>i</u>					

	Conti	nuous Assess	CAM)	End			
Assessment	Assignment 1	Assignment 2	Test 1	Test 2	Attendance	Semester Examination (ESE) Marks	Total Marks
Marks	20	20	25	25	10	1	100

5. V) / · ·

Department	Artific	ial Intelli	gence and Da	ata Science	Program	nme: M	.Tech.				
Semester	1/11				Course	Catego	ry : AE (C *End So	emester Ex	am Type: -	
Course Code	DOSA	CTX08			Perio	ds / W	eek	Credit	t Ma	rks	
Course Code	FZSA	CIAUO			L	Т	Р	С	CAM	ESE	TM
Course Name	1	UGH LIFI	DEVELOPM E ENLIGHTEI		0	0	2	-	100	-	100
				(Commo	on to all M	.Tech F	Program	ıme)			
Prerequisite	No Pı	rerequisite	e needed								
	On co	On completion of the course, the students will be able to (Hig									
Course	CO1 Study of Shrimad-Bhagwad-Geeta will help the student in developing his personality and achieve the highest goal in life.										3
Outcomes	CO2	prosperity.									
	CO3	Study of I	Neet is hatakan	n will help in de	veloping ve	rsatile p	ersonali	ty of studen	nts.	K3	3
UNIT- I								Periods:	: 6		
Neetisatakam-hol 26,28,63,65 (virtu											CO1
UNIT- II								Periods:	: 12		
Approach to day t 35 Chapter 6-Ver Chapter 3-Verses	ses 5,13	3,17,23, 35	- Chapter 18-\								CO2
UNIT- III								Periods:	: 12		
Statements of bas - Personality of ro		rledge – Sl	nrimad Bhagwa	d Geeta: Chap	ter2-Verses	56, 62	, 68 Cha	pter12 -Ver	ses 13, 14, 1	5, 16,17, 18	CO3
Lecture Perio	ds: 30		Tutorial Per	iods: -	Practica	al Perio	ods: -		Total Peri	ods: 30	
Reference Boo	ks	······						<u></u>			
1. Gopinath, Rasł 2. Swami Swarup										-	

	Conti	nuous Assess	ment N	larks (0	CAM)	End	
Assessment	Assignment 1	Assignment 2	Test 1	Test 2	Attendance	Semester Examination (ESE) Marks	Total Marks
Marks	20	20	25	25	10	-	100

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Department	Artifici	Artificial Intelligence and Data Science Programme: M.Tech.								
Semester	1/11			Course	Catego	ory : AE	C *End \$	nd Semester Exam Type: -		
Course Code	Ρ23Δ	CTX09		Perio	ods / W	eek	Credi	t Ma	Maximum Mar	
Oddisc Oddc				L	Т	Р	С	CAM	ESE	TM
Course Name	UNNA	AT BHAR	ATH ABHIYAN	0	0	2	-	100	-	100
			······································	on to all M	.Tech I	Progran	nme)			
Prerequisite	Į		te needed							
		-	n of the course, the stude						BT Mappi (Highest Le	
	CO1	Gain an ι	inderstanding of rural life, cultu	re and soc	ial reali	ties			K	3
Course Outcomes	CO2	Develop a	a sense of empathy and bonds	of mutuali	ty with l	ocal com	munity		K	1
Odtoomes	CO3	CO3 Appreciate significant contributions of local communities to Indian society and economy								
	CO4 Learn to value the local knowledge and wisdom of the community									3
	CO5	CO5 Identify opportunities for contributing to community's socio-economic improvements.								
UNIT- I	Appre	ciation o	f Rural Society				Periods	: 6		
			and gender relations, rural valu illages' (Gandhi), rural infrastru		spect to	commur	nity, nature a	and resource	S,	CO1
UNIT- II	Unde	rstandin	g Rural Economy and Live	elihood			Periods	: 6		
entrepreneurs, rui	al mark	ets.	ip, water management, ani	mal husb	andry,	non-farr	·•		sans, rural	CO
UNIT- III		Instituti					Periods			
Fraditional rural Committees), loca			elf-help Groups, Panchayati al administration.	raj institut	tions (C	Gram Sa	abha, Gran	n Panchayat	, Standing	CO
UNIT- IV	Rural	Develop	oment Programmes				Periods	: 6		
			, current national programmes: Yojana, Skill India, Gram Pan							CO4
UNIT- V	Field	Based P	ractical Activities				Periods	: 6		
vith local leaders	s, Éanch	ayat func	thh Bharat project sites, Condu tionaries, Visit Rural Schools Gram Sabha meetings, Visit loo	/ mid-day	meal c	entres,	study Acad	emic and inf	rastructural	CO
Lecture Perio	ds: 30		Tutorial Periods: -	Practic	al Peri	ods: -		Total Peri	ods: 30	<u>i</u>
Reference Boo	ks						<u></u>			
2. A Hand book of B. United Nations,	n Village "Sustai	Panchay nable Dev	t : Principles, Policies and Mar at Administration, Rajiv Gandh elopment Goals", 2015. Bural Development", Shanlax P	i Chair for	Pancha					

	Conti	nuous Assess	CAM)	End			
Assessment	Assignment 1	Assignment 2	Test 1	Test 2	Attendance	Semester Examination (ESE) Marks	Total Marks
Marks	20	20	25	25	10	-	100

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Department	Artificial Intelligence and Data Science	Program	nme: M	.Tech.					
Semester	III	Course	Catego	ry : PA	*End Semester Exam Type: LE				
Course Code	P23ADW301	Perio	ds / We	eek	Credit	Maximum Marks			
Course Code	1 23AD 11301	L	Т	Р	С	CAM	ESE	TM	
Course Name	Project Phase - I	0	0	12	6	50	50	100	

Aim & Objective:

The project work aims to develop the work practice and to apply theoretical and practical tools/techniques for solving real life problems related to industry and current research. The objective of the project work is to improve the professional competency and research attitude by touching the areas which are not covered in theory or laboratory classes.

- The project work shall be a design project/experimental project and/or computer simulation project on any of the topic in manufacturing engineering or related field.
- The project work shall be allotted individually on different topics.
- The students shall be encouraged to do their project work in the parent institute itself. In exceptional cases the students shall be permitted to undertake continue their project outside the parent institute with appropriate permission from Head of the institution through the Project Coordinator.
- Department shall constitute an Evaluation Committee to review the project work.
- The Evaluation committee shall consist of at least three faculty members namely internal guide, project coordinator and another expert in the specified area of the project.

The student is required to undertake the project phase I during the third semester and the same shall be continued in the 4 th semester (Phase II). Phase I consist of preliminary thesis work, three reviews of the work and the submission of preliminary report. First review shall highlight the topic, objectives and origin of problem, second review shall highlight, Literature survey, methodology and expected results. Third review shall evaluate the progress of the work, preliminary report and scope of the work which shall be completed in the 4 th semester. Also the evaluation of project phase - I shall be done externally.

Department	Artificial Intelligence and Data Science	Program	nme: M	Tech.				
Semester	III	Course Category : PA *End Semester Exam Ty						Type: -
Course Code	P23ADW302	Perio	ds / We	eek	Credit	Maximum Marks		
Course Code		L	Т	Р	С	CAM	ESE	TM
Course Name	Internship	0	0	0	2	100	-	100

Students should undergo training or internship during summer / winter vacation at Industry/ Research organization / University (after due approval from the Programme Academic Coordinator and Department Consultative Committee (DCC). In such cases, the internship/training should be undergone continuously (without break) in one organization. Normally no extension of time is allowed. However, DCC may provide relaxation based on the exceptional case. The students are allowed to undergo three to four weeks internship in established industry / Esteemed institution during vacation period. The student should give presentation and submit report to DCC. The Internship is assessed internally for 100 marks.

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Department	Artificial Intelligence and Data Science	Programme: M.Tech.							
Semester	III	Course Category : AEC *End Semester Exam Type							
Course Code	P22ADC201	Perio	ds / We	eek	Credit Maxim		ximum M	arks	
Course Code	1 23ADC301	L	Т	Р	С	CAM	ESE	TM	
Course Name	NPTEL/SWAYAM/MOOC	0	0	0	-	100	-	100	

Student should register online courses like MOOC / SWAYAM / NPTEL etc. approved by the Department committee comprising of HoD, Programme Academic Coordinator and Subject Experts. Students have to complete relevant online courses successfully. The list of online courses is to be approved by Academic Council on the recommendation of HoD at the beginning of the semester if necessary, subject to ratification in the next Academic council meeting. The Committee will monitor the progress of the student and recommend the grade (100% Continuous Assessment pattern) based on the marks secured in online examinations. The marks attained for this course is not considered for CGPA calculation.

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Department	Artificial Intelligence and Data Science	Programme: M.Tech.						
Semester	IV	Course Category: PA			*End Semester Exam Type: LE			
Course Code	P23ADW403	Periods / Week			Credit	Maximum Marks		
		L	Т	Р	С	CAM	ESE	TM
Course Name	Project Phase - II	0	0	24	12	50	50	100
	<u>I</u>							

Aim & Objective:

The project work aims to develop the work practice and to apply theoretical and practical tools/techniques for solving real life problems related to industry and current research. The objective of the project work is to improve the professional competency and research attitude by touching the areas which are not covered in theory or laboratory classes.

- The project work shall be a design project/experimental project and/or computer simulation project on any of the topic in manufacturing engineering or related field.
- The project work shall be allotted individually on different topics.
- The students shall be encouraged to do their project work in the parent institute itself. In exceptional cases the students shall be permitted to undertake continue their project outside the parent institute with appropriate permission from Head of the institution through the Project Coordinator.
- Department shall constitute an Evaluation Committee to review the project work.
- The Evaluation committee shall consist of at least three faculty members namely internal guide, project coordinator and another expert in the specified area of the project.

Project phase II is a continuation of project phase I which started in the third semester. There shall be three reviews in the fourth semester, first in the beginning of the semester, second in the middle of the semester and the Third at the end of the semester. First review is to evaluate the progress of the work and planned activity; second review shall be presentation and discussion. Third review shall be a presubmission presentation before the evaluation committee to assess the quality and quantity of the work done. This would be a pre qualifying exercise for the students for getting approval for the submission of the thesis. At least one technical paper shall be prepared for possible publication in journals or conferences. The technical paper shall be submitted along with the thesis. The final evaluation of the project shall be done externally.

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