



SRI MANAKULA VINAYAGAR
ENGINEERING COLLEGE
(AN AUTONOMOUS INSTITUTION)



Fourth Meeting of the Board of Studies

Department of Computational Studies

for the Programme

Bachelor of Data Science and Analytics

Venue

First Floor, SAS Block

Sri Manakula Vinayagar Engineering College

Madagadipet, Puducherry – 605 107

Date & Time

10-04-2024 & 10.00 pm to 01.00 pm

5/4

R.D. Mohanalingam



School of Arts and Science
Department of Computational Studies
Board of Studies Meeting for B.Sc. Data Science and Analytics

The fourth meeting of Board of Studies for the course B.Sc. Data Science and Analytics was held through online on 10.04.2024 at 10:00 am in the Department of Computational Studies, School of Arts and Science, Sri Manakula Vinayagar Engineering College with the Head of the Department in the Chair.

The following members were present for the Fourth Meeting of Board of Studies.

S. No.	Name of the Member with Designation and official Address	Responsibility in the BoS
1	Dr. N. MOGANARANGAN, M.E., Ph.D. Professor & Head, Department of Computational Studies, School of Arts & Science, Sri Manakula Vinayagar Engineering College (Autonomous) Madagadipet, Puducherry 605 107 E-mail: moganarangan.cse@smvec.ac.in Mobile: 98945 33661	Chairman
2	Dr. PUNAM BEDI M.Sc., M.Tech., Ph.D. Professor, Department of Computer Science, University of Delhi, Delhi – 110 007. Email: punambedi@gmail.com , Mobile:9899125785	Pondicherry University Nominee
3	Dr. R. AROKIA PAUL RAJAN MCA, PGDBA, MA, Ph.D., Associate Professor, Computer Science, School of Sciences, Bangalore Central Campus, Christ University, Bangalore, Karnataka. Mail id: paulraajan@gmail.com, Ph: 9443459242	Subject Expert (Academic Council Nominee)
4	Dr. M. DURAISAMY, MCA., M.Phil., Ph.D., TNSET. Associate Professor and Head, Department of Computer Applications, Government Arts and Science College, Kariyampatti, Tirupattur, Tamilnadu - 635 901. E-mail: duraimca78@gmail.com, Mobile: 98431 55358.	Subject Expert (Academic Council Nominee)
5	Mr.E.IYYAPPAN, Senior Application Developer, IQVIA, Bangalore. E-mail: eiyvappan.mca@gmail.com, Mobile:9790700670	Member (Industry Representative)

5/4

R.D. Mohanalingam

Co-opted Expert Members		
6	<p>Dr. J. MADHUSUDANAN, ME., Ph.D., Professor and Head, Department of Artificial Intelligence and Data Science, Sri Manakula Vinayagar Engineering College, Madagadipet, Puducherry. 605 107. E-mail: madhu@smvec.ac.in Mobile: 9003739274</p>	Co-opted Expert Member
7	<p>Mr. M. SHANMUGAM, M.Sc., M.Phil., M.E., SET, (Ph.D.), Associate Professor, Department of Computer Science Engineering, Sri Manakula Vinayagar Engineering College E-mail: shanmugam.mm@smvec.ac.in Mobile: 9444370963</p>	Co-opted Expert Member
Internal Members		
8	<p>Mrs. A. SHAMSATH BEGAM, M.C.A., Assistant Professor, Department of Computational Studies, School of Arts and Science, Sri Manakula Vinayagar Engineering College, Madagadipet, Puducherry. 605 107. E-mail: shamsathbegum.sas@smvec.ac.in, Mobile: 9500399774</p>	Internal Member
9	<p>Mr. K. SANTHOSHKUMAR, M.C.A. M.Phil. B.Ed., Assistant Professor Department of Computational Studies School of Arts and Science Sri Manakula Vinayagar Engineering College Mail id: santhoshkumark.sas@smvec.ac.in, Phone: 8508068040.</p>	Internal Member
10	<p>Dr. M.A. ISHRATH JAHAN M.A., M.Phil., Ph.D., Associate Professor & Head, Department of English, School of Arts and Science, Sri Manakula Vinayagar Engineering College, Madagadipet, Puducherry. 605 107. E-mail: ishrath@smvec.ac.in, Mobile: 9443075126.</p>	Internal Member
11	<p>Mr. P.KRISHNAMOORTHY M.Sc., M.Phil., Assistant Professor and Head, Department of mathematics, School of Arts and Science, Sri Manakula Vinayagar Engineering College, Madagadipet, Puducherry. 605 107. E-mail: krishnamoorthymaths@smvec.ac.in, Mobile: 9750028056.</p>	Internal Member

ANNEXURE – II

AGENDA OF THE MEETING

Item No.: BoS//2024/SAS/DSA / 4.1

- ❖ Confirmation of Minutes of 3rd Meeting of BOS - Modifications if any.

Item No.: BoS//2024/SAS/DSA / 4.2

- ❖ To discuss the criteria for fixing (MJD) Major Disciplinary Courses, (MID) Minor Disciplinary Courses, (MLD) Multi-Disciplinary Courses, (AEC) Ability Enhancement Courses, (SEC) Skill Enhancement Courses and (VAC) Value Added Courses from semesters II-VIII.

Item No.: BoS//2024/SAS/DSA / 4.3

- ❖ To discuss the curriculum framework for the UG programme B.Sc. Data Science and Analytics for semesters (I – VIII) based on National Education Policy Regulations 2023 of Pondicherry University.

Item No.: BoS//2024/SAS/DSA / 4.4

- ❖ To discuss the already approved curriculum R2020 and the syllabus of Semester V and Semester VI.
- ❖ To add Value Added Course – Understanding India in II semester.

Item No.: BoS//2024/SAS/DSA / 4.5

- ❖ Any other additional points to be discussed with the permission of the chair.





SRI MANAKULA VINAYAGAR
ENGINEERING COLLEGE
(AN AUTONOMOUS INSTITUTION)



SCHOOL OF ARTS AND SCIENCE

Department of Computational Studies

B.Sc. Data Science and Analytics

Minutes of 4th meeting of Board of Studies

ANNEXURE – I

ACADEMIC REGULATIONS 2020

(R 2020)

5/11

R.D. Mohanalingam

COLLEGE VISION AND MISSION

Vision

To be globally recognized for excellence in quality education, innovation and research for the transformation of livestock 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 instil deep sense of human values by blending societal righteousness with academic professionalism for the growth of society.

DEPARTMENT OF COMPUTATIONAL STUDIES

VISION AND MISSION

Vision:

To come up with successfully as a high-quality human capital in Computer Science and related areas for the sustainable growth of the IT industry needs of the country.

Mission:

M1: Innovative Skills:

Ensuring deeper understanding of fundamentals and acquiring innovative skills within core areas of Computer Science.

M2: Motivated Graduates:

Producing highly skilled and motivated graduates with the ability of problem solving individually and in teams.

M3: Ethical Responsibilities:

Providing a deep awareness of our ethical responsibilities to our profession and to the society.



SEMESTER - V

S. No	Course Code	Course Title	Category	Periods			Credits	Max. Marks		
				L	T	P		CAM	ESM	Total
Theory										
1	A20DAT509	Introduction to Machine Learning	DSC	4	0	0	4	25	75	100
2	A20DAT510	IoT Cloud and Data Analytics	DSC	4	0	0	4	25	75	100
3	A20DAT511	Software Project Management	DSC	4	0	0	4	25	75	100
4	A20DAE5XX	Discipline Specific Elective-III	DSE	3	0	0	3	25	75	100
Practical										
6	A20DAL509	Machine Learning Lab	DSC	0	0	4	2	50	50	100
7	A20DAP501	Mini Project	DSC	0	0	4	2	50	50	100
Skill Enhancement Course										
8	A20DAS505	R Programming Lab	SEC	0	0	4	2	100	0	100
Other Certification Course										
9	A20DAX501	NPTEL – Big Data Computing	OCC	0	0	0	0	0	0	100
							21	300	400	800

R.D. Mohanalingar

DISCIPLINE SPECIFIC ELECTIVES

Sl. No	Course Code	Course Title	Category	Periods			Credits	Max. Marks		
				L	T	P		CA M	ES M	Total
Discipline Specific Elective (DSE - I) – offered in Third Semester										
1	A20DAE301	Operating System	DSE	3	-	-	3	25	75	100
2	A20DAE302	Information Security	DSE	3	-	-	3	25	75	100
3	A20DAE303	Computer Networks	DSE	3	-	-	3	25	75	100
Discipline Specific Elective (DSE - II) – offered in Fourth Semester										
1	A20DAE404	Infrastructure Management	DSE	3	-	-	3	25	75	100
2	A20DAE405	Client Server Technology	DSE	3	-	-	3	25	75	100
3	A20DAE406	Image Processing	DSE	3	-	-	3	25	75	100
Discipline Specific Elective (DSE - III) – offered in Fifth Semester										
1	A20DAE507	Wireless Sensor Network	DSE	3	-	-	3	25	75	100
2	A20DAE508	Data Science using R	DSE	3	-	-	3	25	75	100
3	A20DAE509	Virtualization using Cloud	DSE	3	-	-	3	25	75	100
Discipline Specific Elective (DSE - IV) – offered in Sixth Semester										
1	A20DAE610	Process Management	DSE	3	-	-	3	25	75	100
2	A20DAE611	Software Engineering	DSE	3	-	-	3	25	75	100
3	A20DAE612	Introduction to Digital Marketing	DSE	3	-	-	3	25	75	100



R.D. Mohanalingar

Department	Computational Studies			Programme: B.Sc DATA SCIENCE AND ANALYTICS						
Semester	Fifth			Course Category Code: DSC *End Semester Exam Type: TE						
Course Code	A20DAT509			Periods / Week			Credit	Maximum Marks		
				L	T	P	C	CAM	ESE	TM
Course Name	INTRODUCTION TO MACHINE LEARNING			4	0	0	4	25	75	100
Prerequisite	Basic knowledge about Machine Learning									
Course Objectives	<ul style="list-style-type: none"> To understand the machine learning theory. To implement linear and non-linear learning models. To implement distance-based clustering techniques. To build tree and rule based models. To apply reinforcement learning techniques 									
Course Outcome	<i>After the completion of this course, the students will be able to:</i>							BT Mapping (Highest Level)		
	CO1	Understanding the Machine Learning Concepts						K2		
	CO2	Implement the Machine Learning Models						K3		
	CO3	Analyse the principle of cluster technologies						K3		
	CO4	Implement the linear and Boolean Functions						K4		
CO5	Methods to work with Kernal Principles						K4			
UNIT-I	INTRODUCTION TO MACHINE LEARNING					Periods: 12				
Introduction – machine learning applications – learning associations, classification, regression, unsupervised learning – Reinforcement Learning – Supervised Learning – VC Dimension – PAC learning – noise – learning multiple classes – regression – model selection and generalization – Bayesian Decision Theory – Classification – losses and risks – Discriminant Functions.								CO1		
UNIT-II	MACHINE LEARNING MODELS					Periods: 12				
Bernoulli Density – Multinomial Density – Gaussian Density – Bias – Variance – Parametric classification – Multivariate Methods – Multivariate Data – Parameter Estimation – Estimation of Missing Values – Multivariate Normal Distribution – Multivariate Classification – Multivariate Regression - Dimensionality Reduction – Subset Selection – Principal Component Analysis – Factor Analysis – Dimensionality Scaling – Linear Discriminant Analysis.								CO2		
UNIT-III	CLUSTER TECHNOLOGIES					Periods: 12				
Clustering – Mixture Densities – K Means Clustering – Expectation Maximization – Mixtures of Latent Variable Models – Supervised Learning after clustering – Hierarchical Clustering – Non-Parametric Methods – Histogram Estimator – Kernel Estimator – K-Nearest Neighbor Estimator.								CO3		
UNIT-IV	LINEAR AND BOOLEAN FUNCTIONS					Periods: 12				
Linear Discrimination – Linear Model – Geometry of the Linear Discriminant – Pairwise Separation – Gradient Descent – Logistic Discrimination - Introduction – Perceptron – Training a Perceptron – Learning Boolean Functions – Multilayer Perceptrons – Backpropagation Algorithm								CO4		
UNIT-V	KERNAL ANALYSIS					Periods: 12				
Kernel Machines – Optimal Separating Hyperplane – kernel trick – vectorial kernels – multiple kernel learning – multiclass kernel machines – Hidden Markov Models - Finding the State Sequence – Learning Model Parameters – Generating Diverse Learners – Model Combination schemes – Voting – Error Correcting Output Codes – Bagging – Boosting								CO5		
Lecture Periods: 60			Tutorial Periods:			Practical Periods: -			Total Periods: 60	
			-							
Text Books										
1. EthemAlpaydin, Introduction to Machine Learning, Second Edition, PHI,2010.										
2. Machine Learning, An Algorithmic Perspective, Stephen Marsland, Chapman & Hall Machine Learning and Pattern Recognition Series, Second Edition, CRC Press,2009.										
Reference Books										
1. Machine Learning, The Art and Science of Algorithms that Make Sense of Data, Peter Flach,										

Cambridge University Press.

2. Machine Learning, Tom M. Mitchell, Mc Graw Hill, 2003

Web References

1. <https://www.geeksforgeeks.org/machine-learning/>
2. <https://www.ibm.com/topics/machine-learning>
3. <https://www.coursera.org/articles/what-is-machine-learning>
4. <https://developers.google.com/machine-learning/crash-course/>
5. <https://cloud.google.com/learn/what-is-machine-learning>

** TE – Theory Exam, LE – Lab Exam

COs/POs/PSOs Mapping

COs	Program Outcomes (POs)					Program Specific Outcomes (PSOs)		
	PO1	PO2	PO3	PO4	PO5	PSO1	PSO2	PSO3
1	2	2	3	3	3	2	2	2
2	2	3	2	3	2	3	2	2
3	3	2	3	3	2	3	3	2
4	3	2	2	2	2	3	2	3
5	2	2	2	3	3	3	2	2

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

Evaluation Method

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

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

Department	Computational Studies		Programme: B.Sc DATA SCIENCE AND ANALYTICS							
Semester	Fifth		Course Category Code: DSC			*End Semester Exam Type: LE				
Course Code	A20DAT510		Periods / Week			Credit	Maximum Marks			
			L	T	P	C	CAM	ESE	TM	
Course Name	IOT Cloud and Data Analytics		4	0	0	4	25	75	100	
Prerequisite	Basic knowledge in IOT Cloud and Data moduels									
Course Objectives	<ul style="list-style-type: none"> To understand IOT theory. To implement Data Analytics models. To implement Web of things techniques. To build various network. To apply cloud techniques to create real time applications. 									
Course Outcome	<i>After the completion of this course, the students will be able to:</i>							BT Mapping (Highest Level)		
	CO1	To impart knowledge on the Internet of Things						K2		
	CO2	To analyze, design and develop IoT solutions						K3		
	CO3	To apply the concept of IOT in the real world apps.						K3		
	CO4	To use various techniques of data storage and analytics in IoT						K4		
	CO5	To understand APIs to connect IoT related technologies						K4		
UNIT-I	INTRODUCTION TO FUNDAMENTALS OF IOT					Periods: 12				
Introduction - Definitions & Characteristics of IoT - IoT Architectures - Physical & Logical Design of IoT - Enabling Technologies in IoT - History of IoT - About Things in IoT - The Identifiers in IoT About the Internet in IoT - IoT frameworks - IoT and M2M.								CO1		
UNIT-II	DATA ANALYTICS					Periods: 12				
Introduction to elastic analytics - Decouple key components - Cloud security and analytics - Designing data processing for analytics - Applying big data technology to Storage								CO2		
UNIT-III	IOT TO WOT					Periods: 12				
The Future Web of Things – Set up cloud environment –Cloud access from sensors- Rest Architectures- The web of Things Resource Identification - Identifier Richardson Maturity Model								CO3		
UNIT-IV	IOT PROTOCOLS AND NETWORKS					Periods: 12				
Control Units – Communication modules – Bluetooth – Zigbee – Wifi – GPS- IOT Protocols (IPv6, 6LoWPAN, RPL, CoAP etc..) – MQTT - Wired Communication - Power Sources								CO4		
UNIT-V	APPLICATIONS OF IOT WITH CLOUD					Periods: 12				
Home Automation - Smart Cities – Energy - Retail Management – Logistics – Agriculture - Health and Lifestyle - Industrial IoT - Legal challenges - IoT design Ethics - IoT in Environmental Protection.								CO5		
Lecture Periods: 60			Tutorial Periods:			Practical Periods: -			Total Periods: 60	
			-							
Text Books										
1. J. Biron and J. Follett, "Foundational Elements of an IoT Solution", O'Reilly Media, 2016.										
2. Simone Cirani, Gianluigi Ferrari, Marco Picone, Luca Veltri. Internet of Things: Architectures, Protocols and Standards, 1 st edition, Wiley Publications, 2019.										
3. Olivier Hersent, David Boswarthick, and Omar Elloumi, — “The Internet of Things: Key Applications and Protocols”, WileyPublications										

Reference Books

1. Vermesan, Ovidiu, and Peter Friess, eds. Internet of things-from research and innovation to market deployment, 1st edition, Aalborg: River publishers, 2014.
2. Daniel Minoli, — “Building the Internet of Things with IPv6 and MIPv6: The Evolving World of M2M Communications”, ISBN: 978-1-118-47347-4, Willy Publications
3. Pethuru Raj and Anupama C. Raman, "The Internet of Things: Enabling Technologies, Platforms, and Use Cases"

Web References

1. <https://www.geeksforgeeks.org/introduction-to-internet-of-things-iot-set-1/>
2. <https://cloud.google.com/discover/what-is-cloud-analytics>
3. <https://www.javatpoint.com/iot-internet-of-things>
4. <https://developer.ibm.com/articles/cl-cloudintro/>
5. <https://www.tutorialspoint.com/iot-network-protocols>

* TE – Theory Exam, LE – Lab Exam

COs/POs/PSOs Mapping

COs	Program Outcomes (POs)						Program Specific Outcomes (PSOs)		
	PO1	PO2	PO3	PO4	PO5	PO6	PSO1	PSO2	PSO3
1	3	2	3	3	2	2	2	2	3
2	3	2	2	3	3	3	2	2	2
3	2	2	2	2	3	3	3	2	2
4	3	3	2	2	2	2	2	2	3
5	2	3	2	2	3	3	3	3	3

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

Evaluation Method

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

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

R.D. Mithal

Department	Computational Studies		Programme: B.Sc Data Science and Analytics						
Semester	Fifth		Course Category Code: DSC			*End Semester Exam Type: TE			
Course Code	A20DAT511		Periods / Week			Credit	Maximum Marks		
			L	T	P	C	CAM	ESE	TM
Course Name	Software Project Management		4	0	0	4	25	75	100
Prerequisite									
Course Objectives	<ul style="list-style-type: none"> To understand the machine learning theory. To implement linear and non-linear learning models. To implement distance-based clustering techniques. To build tree and rule based models. 								
Course Outcome	After the completion of this course, the students will be able to:							BT Mapping (Highest Level)	
	CO1	Understand the Software Project Management concepts.						K2	
	CO2	Design and apply cost estimation						K3	
	CO3	Understand and evaluate software quality management						K3	
	CO4	Design and create software management and metrics						K4	
	CO5	Analyze the role of project evaluation						K4	
UNIT-I	Project Concepts and Its Management						Periods: 12		
Project life cycle models-ISO 9001 model - Capability Maturity Model - Project, Planning-Project tracking-Project closure - Evolution of Software Economics – Software Management Process Framework: Phases, Artifacts, Workflows, Checkpoints – Software Management Disciplines: Planning / Project Organization and Responsibilities .							CO1		
UNIT-II	Cost Estimation						Periods: 12		
Problems in Software Estimation – Algorithmic Cost Estimation Process, Function, Points, SLIM (Software Life cycle Management), COCOMO II (Constructive Cost Model) – Estimating Web Application Development – Concepts of Finance, Activity Based Costing and Economic Value Added (EVA) – Balanced Score Card.							CO2		
UNIT-III	Software Quality Management						Periods: 12		
Software Quality Factors – Software Quality Components – Software Quality Plan– Software Quality Metrics – Software Quality Costs – Software Quality Assurance-Standard – Certification – Assessment.							CO3		
UNIT-IV	Software Management and Metrics						Periods: 12		
Software Configuration Management – Risk Management: Risk Assessment: Identification / Analysis / Prioritization – Risk Control: Planning / Resolution /Monitoring – Failure Mode and Effects Analysis (FMEA) –Defect Management-Cost Management. Software Metrics – Classification of Software Metrics.							CO4		
UNIT-V	Project Evaluation and Emerging Trends						Periods: 12		
Strategic Assessment–Technical Assessment–Cost Benefit Analysis–Cash Flow Forecasting–Cost Benefit Evaluation Technique–Risk Evaluation–Software Effort Estimation. Emerging Trends: Import of the internet on project Management –people Focused Process Models.							CO5		
Lecture Periods: 60			Tutorial Periods: -			Practical Periods: -		Total Periods: 60	

Text Books

1. Ramesh Gopaldaswamy, Managing and Global Software Projects, Tata McGraw Hill, 2017.
2. Neal Whitten, Managing Software Development Projects, John Wiley & Sons, Inc., 2nd Ed., 1995.
3. Kaplan, R.S., Norton, D.P. The Balanced Scorecard: Translating Strategy into Action, Harvard Business School Press, 1996.
4. Boehm, B. W. Software Risk Management: Principles and Practices in IEEE Software, January 1991, pp32-41. Grant, J.L. Foundations of Economic Value Added, John Wiley & Sons, 1997.

Reference Books

1. Demarco, T. and Lister, T. Peopleware: Productive Projects and Teams, 2nd Ed., Dorset House, 1999.
2. Royce, W. Software Project Management: A Unified Framework, Addison-Wesley, 1998. Demarco, T. and Lister, T. Peopleware: Productive Projects and Teams, 2ndEd., Dorset House, 1999.
3. Fenton, N.E., and Pfleeger, S.L. Software Metrics: A Rigorous and Practical Approach, Revised Brooks Cole, 1998.

* TE – Theory Exam, LE – Lab Exam

COs/POs/PSOs Mapping

COs	Program Outcomes (POs)					Program Specific Outcomes (PSOs)		
	PO1	PO2	PO3	PO4	PO5	PSO1	PSO2	PSO3
1	2	2	3	3	3	2	2	2
2	2	3	2	3	2	3	2	2
3	3	2	3	3	2	3	3	2
4	3	2	2	2	2	3	2	3
5	2	2	2	3	3	3	2	2

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

Evaluation Method

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

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

Department	Computational Studies			Programme: B.Sc DATA SCIENCE AND ANALYTICS							
Semester	Fifth			Course Category Code: DSC *End Semester Exam Type: TE							
Course Code	A20DAE507			Periods / Week		Credit	Maximum Marks				
				L	T	P	C	CAM	ESE	TM	
Course Name	WIRELESS SENSOR NETWORK			3	0	0	3	25	75	100	
Prerequisite	Basic knowledge about wireless sensor network										
Course Objectives	<ul style="list-style-type: none"> To understand the wireless sensors network To implement WSN communication To implement MAC and routing protocol To understand network topology. To apply reinforcement learning techniques 										
Course Outcome	After the completion of this course, the students will be able to:									BT Mapping (Highest Level)	
	CO1	To understand the fundamentals of wireless sensor network								K2	
	CO2	To gain knowledge on the MAC and Routing Protocols of WSN								K3	
	CO3	To get exposed to networking sensors technology								K3	
	CO4	To acquire knowledge on the protocols required for developing real time applications using network topology								K4	
	CO5	To acquire sensor network platforms								K4	
UNIT-I	OVERVIEW OF WIRELESS SENSOR NETWORKS							Periods: 9			
Introduction Single-Node Architecture - Hardware Components - Network Characteristics-unique constraints and challenges, Enabling Technologies for Wireless Sensor Networks - Types of wireless sensor networks.										CO1	
UNIT-II	ARCHITECTURES							Periods: 9			
Network Architecture - Sensor Networks-Scenarios - Design Principle, Physical Layer and Transceiver Design Considerations, Optimization Goals and Figures of Merit, Gateway Concepts, Operating Systems and Execution Environments - introduction to Tiny OS and C - Internet to WSN Communication.										CO2	
UNIT-III	NETWORKING SENSORS							Periods: 9			
MAC Protocols for Wireless Sensor Networks, Low Duty Cycle Protocols And Wakeup Concepts – SMAC, - B-MAC Protocol, IEEE 802.15.4 standard and ZigBee, the Mediation Device Protocol, Wakeup Radio Concepts, Address and Name Management, Assignment of MAC Addresses, Routing Protocols Energy-Efficient Routing, Geographic Routing										CO3	
UNIT-IV	INFRASTRUCTURE ESTABLISHMENT							Periods: 9			
Topology Control, Clustering, Time Synchronization, Localization and Positioning, Sensor Tasking and Control.										CO4	
UNIT-V	SENSOR NETWORK PLATFORMS AND TOOLS							Periods: 9			
Sensor Node Hardware – Berkeley Motes, Programming Challenges, Node-level software platforms, Node-level Simulators, State-centric programming.										CO5	
Lecture Periods: 45			Tutorial Periods:			Practical Periods: -			Total Periods: 45		
Text Books											
1.Holger Karl & Andreas Willig, "Protocols And Architectures for Wireless Sensor Networks", John											

R.D. Mithunalingar

Wiley, 2005.

2. Feng Zhao & Leonidas J. Guibas, "Wireless Sensor Networks- An Information Processing Approach", Elsevier, 2007.

3. Walteneus Dargie, Christian Poellabauer, "Fundamentals of Wireless Sensor Networks - Theory and Practice", John Wiley & Sons Publications, 2011

Reference Books

1. Kazem Sohraby, Daniel Minoli, & Taieb Znati, "Wireless Sensor Networks-Technology, Protocols, and Applications", John Wiley, 2007.

2. Anna Hac, "Wireless Sensor Network Designs", John Wiley, 2003

Web References

6. <https://www.geeksforgeeks.org/wireless-sensor-network/>

7. <https://www.ibm.com/Wireless-sensor-network/>

8. <https://www.coursera.org/articles/what-is-wireless-sensor-network>

* TE – Theory Exam, LE – Lab Exam

COs/POs/PSOs Mapping

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

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

Assessment Pattern as per Bloom's Taxonomy

Test / Bloom's Category*	Remembering (K1) %	Understanding (K2) %	Applying (K3) %	Analyzing (K4) %	Evaluating (K5) %	Creating (K6) %	Total %
CAT1	10	20	70	-	-	-	100
CAT2	10	20	70	-	-	-	100
ESE	10	30	60	-	-	-	100

* ±3% may be varied

Evaluation Method

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

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

R.D. Mithal

Department	Computational Studies		Programme: B.Sc DATA SCIENCE AND ANALYTICS						
Semester	Fifth		Course Category Code: DSE *End Semester Exam Type: TE						
Course Code	A20DAE508		Periods / Week			Credit	Maximum Marks		
			L	T	P	C	CAM	ESE	TM
Course Name	Data Science using R		3	0	0	3	25	75	100
Prerequisite	Basic knowledge in Data Science using R								
Course Objectives	<ul style="list-style-type: none"> To understand the Data Science. To understand the basic of R language. To implement data frame techniques. To implement function and data visualization. To apply inferential statics with R. 								
Course Outcome	<i>After the completion of this course, the students will be able to:</i>								BT Mapping (Highest Level)
	CO1	Recognize various disciplines that contribute to a successful data science effort.							K2
	CO2	Understand the processes of data science							K3
	CO3	Be aware of the challenges that arise in data sciences.							K3
	CO4	Develop and appreciate various techniques for data modelling and mining.							K4
	CO5	Be cognizant of ethical issues in many data science tasks.							K4
UNIT-I	INTRODUCTION TO DATA SCIENCE					Periods: 09			
Introduction- Definition - Data Science in various fields -Types of Data: Numeric – Categorical – Graphical – High Dimensional Data – Classification of digital Data: Structured, Semi-Structured and Un-Structured - Example Applications. Sources of Data: Time Series – Transactional Data – Biological Data – Spatial Data – Social Network Data – Data Evolution.									CO1
UNIT-II	INTRODUCTION TO R					Periods: 09			
Introduction - Features of R - Environment - R Studio. Basics of R-Assignment - Modes - Operators - special numbers - Logical values - Basic Functions - R help functions - R Data Structures - Control Structures. Vectors: Definition- Declaration - Generating - Indexing - Naming - Adding & Removing elements - Operations on Vectors - Recycling – Special Operators - Vectorized if- then else-Vector Equality – Functions for vectors - Missing values - NULL values - Filtering & Subsetting.									CO2
UNIT-III	LIST AND DATA FRAMES USING R					Periods: 09			
Matrices - Creating Matrices - Adding or Removing rows/columns - Reshaping - Operations - Special functions on Matrices. Lists - Creating List – General List Operations – Special Functions - Recursive Lists. Data Frames - Creating Data Frames - Naming - Accessing - Adding - Removing - Applying Special functions to Data Frames - Merging Data Frames- Factors and Tables.									CO3
UNIT-IV	DATA VISUALIZATION					Periods: 09			
Functions - Creating User- defined functions - Functions on Function Object - Scope of Variables - Accessing Global, Environment - Closures - Recursion. Data Visualization in R: Types of visualizations - packages for visualizations - Basic Visualizations, Advanced Visualizations and Creating 3D plots.									CO4
UNIT-V	INFERENCE STATISTICS WITH R					Periods: 09			
Types of Learning - Linear Regression- Simple Linear Regression - Implementation in R - functions on lm() - predict() - plotting and fitting regression line. Multiple Linear Regression. Introduction - comparison with simple linear regression - Correlation Matrix - F-Statistic - Target variables Vs Predictors - Identification of significant features - Implementation of Multiple Linear Regression.									CO5
Lecture Periods: 45			Tutorial Periods: -		Practical Periods: -			Total Periods: 45	

Text Books

1. W. N. Venables, D. M. Smith and the R Core Team, "An Introduction to R", 2013.
2. Tony Ojeda, Sean Patrick Murphy, Benjamin Bengfort, Abhijit Dasgupta, "Practical Data Science Cookbook", Packt Publishing Ltd., 2014.
3. Nathan Yau, "Visualize This: The FlowingData Guide to Design, Visualization, and Statistics", Wiley, 2011.

Reference Books

1. Nina Zumel, John Mount, "Practical Data Science with R", Manning Publications, 2014.
2. Jure Leskovec, Anand Rajaraman, Jeffrey D. Ullman, "Mining of Massive Datasets", Cambridge University Press, 2014.
3. Mark Gardener, "Beginning R - The Statistical Programming Language", John Wiley & Sons, Inc., 2012.

Web References

9. <https://www.rdocumentation.org/>
10. <https://www.geeksforgeeks.org/r-programming-for-data-science/>
11. https://onlinecourses.nptel.ac.in/noc21_cs23/preview
12. <https://archive.nptel.ac.in/courses/111/104/111104147/>

* TE – Theory Exam, LE – Lab Exam

COs/POs/PSOs Mapping

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

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

Assessment Pattern as per Bloom's Taxonomy

Test / Bloom's Category*	Remembering (K1) %	Understanding (K2) %	Applying (K3) %	Analyzing (K4) %	Evaluating (K5) %	Creating (K6) %	Total %
CAT1	10	20	70	-	-	-	100
CAT2	10	20	70	-	-	-	100
ESE	10	30	60	-	-	-	100

* ±3% may be varied

Evaluation Method

Assessment	Continuous Assessment Marks (CAM)					End Semester Examination (ESE) Marks	Total Marks
	CAT 1	CAT 2	Model Exam	Assignment *	Attendance		




Marks	10	5	5	5	75	100
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* Application oriented / Problem solving / Design / Analytical in content beyond the syllabus

Department	Computational Studies		Programme: B.Sc DATA SCIENCE AND ANALYTICS						
Semester	Fifth		Course Category Code: DSC *End Semester Exam Type: TE						
Course Code	A20DAE509		Periods / Week			Credit	Maximum Marks		
	L	T	P	C	CAM	ESE	TM		
Course Name	VIRTUALIZATION USING CLOUD		3	0	0	3	25	75	100
Prerequisite	Basic knowledge about virtualization using cloud								
Course Objectives	<ul style="list-style-type: none"> To understand the virtualization concepts To understand virtualization infrastructures. To implement virtualization solution To understand the cloud concepts To apply the concepts of cloud security 								
Course Outcome	After the completion of this course, the students will be able to:							BT Mapping (Highest Level)	
	CO1	To understand the concepts of virtualization and virtual machines.						K2	
	CO2	To gain expertise in server, network and storage virtualization						K3	
	CO3	To gain the knowledge on the concept of virtualization solutions.						K3	
	CO4	To understand the cloud platform architectures.						K4	
	CO5	To understand the security issues in the grid and the cloud environment.						K4	
UNIT-I	VIRTUALIZATION				Periods: 9				
Basic Of Virtual Machines –Process Virtual Machines-System Virtual Machines-Emulation- Interpretation-Binary Translation-Taxonomy Of Virtual Machines.-Virtualization-Management Virtualization-Hardware Maximization-Architectures-Virtualizations Management-Storage Virtualization.								CO1	
UNIT-II	VIRTUALIZATION INFRASTRUCTURE				Periods: 9				
Comprehensive Analysis-Resources Pool-Testing Environment-Server Virtualization-Virtual Workloads-Provision Virtual Machines-Desktop Virtualization-Application Virtualization- Implementation Levels Of Virtualization- Virtualization Structure- Virtualization Of CPU,Memory And I/O Devices-Virtual Clusters And Resources Management – Virtualization For Data Center Automation.								CO2	
UNIT-III	VIRTUALIZATION SOLUTION				Periods: 9				
Understanding Microsofts Virtualization Solutions:Microsoft Infrastructures Optimization Model-Virtualization And The Infrastructure Optimization Model-Benefits Of Virtualization-Achieving The Benefits Of Datacenter Virtualization-Achieving The Benefits Of Client Virtualization-Achieving The Benefits Of Cloud Virtualization.								CO3	
UNIT-IV	CLOUD PLATFORM ARCHITECTURE				Periods: 9				
Cloud Deployment Models:Public,Private,Hybrid,Community-Categories Of Cloud Computing:Everything As A Service:Infrastructures,Platform,Software –A Generic Cloud Architecture Design-Layered Cloud Architectural Development-Virtualization Support And Disaster Recovery-Architectural Design Challenges –Public Cloud Platform:GAE,AWS-Inter-Cloud Resources Management.								CO4	
UNIT-V	CLOUD SECURITY				Periods: 9				
Cloud Infrastructures Security:Network,Os And Application Level-Aspects Of Data Security,Provider Data And Its Security,Identity And Access Management Architecture,IAM Practices In The Cloud,SaaS,Paas,IaaS Availability In The Cloud-Key Privacy Issues In The Cloud								CO5	

-Cloud Security And Trust Management.			
Lecture Periods: 45	Tutorial Periods: -	Practical Periods: -	Total Periods: 45
Text Books			
1.Danielle Ruest, Nelson Ruest , - Virtualization: A Beginner’S Guidel, Mcgrew-III Osborne Media.2009			
2 .Jim Smith,Ravi Nair, ” Virtualization Machines: Versatile Platform For System An Processes”.			
3.John W. Rittinghouse And James F. Ransome.”Cloud Computing: Implementation, Management ,And Security”,CRC Press,2010.			
Reference Books			
1.Publication,2006.Cloud Computing (Principles And Paradigms),Edited by Rajkumar Buyya,James Brobreg Andrzej Goscinski,John Wiley & Sons,Inc.2011.			
2.Cloud Computing A Practical Approaches –Anthony T.Velte,Toby J.Velte Robert Elsenpeter TATA Mcgraw-Hill,New Delhi-2010			
Web References			
13. https://www.geeksforgeeks.org/virtualization using cloud/			
14. https://www.ibm.com/topics/virtualization using cloud			
15. https://www.coursera.org/articles/what-is-cloud security			

* TE – Theory Exam, LE – Lab Exam

COs/POs/PSOs Mapping

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

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

Assessment Pattern as per Bloom’s Taxonomy

Test / Bloom’s Category*	Remembering (K1) %	Understanding (K2) %	Applying (K3) %	Analyzing (K4) %	Evaluating (K5) %	Creating (K6) %	Total %
CAT1	10	20	70	-	-	-	100
CAT2	10	20	70	-	-	-	100
ESE	10	30	60	-	-	-	100

* ±3% may be varied

Evaluation Method

Assessment	Continuous Assessment Marks (CAM)					End Semester Examination (ESE) Marks	Total Marks
	CAT 1	CAT 2	Model Exam	Assignment*	Attendance		

Marks	10	5	5	5	75	100
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* Application oriented / Problem solving / Design / Analytical in content beyond the syllabus.

Department	Computational Studies			Programme: B.Sc DATA SCIENCE AND ANALYTICS						
Semester	Fifth			Course Category Code: DSC		*End Semester Exam Type: LE				
Course Code	A20DAL509			Periods / Week		Credit	Maximum Marks			
				L	T	P	C	IM	ESE	TM
Course Name	MACHINE LEARNING LAB			0	0	4	2	50	50	100
(common to B.Sc (CS) & BCA)										
Prerequisite	Basic Knowledge in Machine Learning									
Course Objectives	<ul style="list-style-type: none"> To Learn the basics of machine learning algorithms. To study about the decision and rule based Learning Concepts. To Learn the Equivalence class transformation Algorithm. To Understand the CSV concepts using python. To study the neural Network concepts. 									
	After completion of the course, the students will be able to									
Course Outcome	CO1	get the knowledge on various Machine Learning								K3
	CO2	to identify the appropriate Machine Learning based on the real-world problem								K3
	CO3	design the applications using various Machine Learning								K3
	CO4	effectively as an individual to understand the concept								K4
	CO5	design the applications using Python								K4
List of Experiment										
<ol style="list-style-type: none"> The probability that it is Friday and that a student is absent is 3 %. Since there are 5 school days in a week, the probability that it is Friday is 20 %. What is that a student is absent given that today is Friday? Apply Baye's rule in python to get the result. Extract the data from database using python Implement k-nearest neighbours classification using python Implement Equivalence Class Transformation Algorithm. Implement linear regression using python. Implement Naive Bayes theorem to classify the English text Implement an algorithm to demonstrate the significance of genetic algorithm Implement and demonstrate the FIND-S algorithm finding the most specific hypothesis based on a given set of training data samples. Read the training data from a .CSV file. Build an Artificial Neural Network by implementing the Backpropagation algorithm and test the same using appropriate data sets. Apply EM algorithm to cluster a set of data stored in a .CSV file. Use the same data set for clustering using k-Means algorithm. 										
Lecture Periods: -			Tutorial Periods: -			Practical Periods: 30		Total Periods: 30		
Text Books										
<ol style="list-style-type: none"> Andreas C. Mueller and Sarah Guido, "Introduction to Machine Learning with Python", O'Reilly Media, Inc. First Edition, 2016. Charu C. Aggarwal "Data Classification algorithms and applications" Chapman & Hall/CRC Data mining and Knowledge series. 										

Reference Books

1. Machine Learning, Tom M. Mitchell, Mc Graw Hill, 2003
2. John Hearty "Advanced Machine Learning with python", Pack Publishing Ltd., 2016.

Web References

1. <https://www.geeksforgeeks.org/machine-learning/>
2. <https://www.ibm.com/topics/machine-learning>
3. <https://www.coursera.org/articles/what-is-machine-learning>
4. <https://developers.google.com/machine-learning/crash-course/>
5. <https://cloud.google.com/learn/what-is-machine-learning>

*LE – Lab Exam

COs/POs/PSOs Mapping

COs	Program Outcomes (POs)					Program Specific Outcomes (PSOs)		
	PO1	PO2	PO3	PO4	PO5	PSO 1	PSO 2	PSO 3
1	3	2	3	3	2	2	2	2
2	2	2	2	3	2	3	2	2
3	2	2	2	2	2	3	3	2
4	3	3	2	2	2	2	2	2
5	3	2	2	2	3	3	3	3

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

Evaluation Method

Assessment	Internal Marks			End Semester Examination (ESE) Marks	Total Marks
	Model Exam	Record	Attendance	50	100
Marks	30	10	10		

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

R.D. Mithal

Department	Computational Studies	Programme: B.Sc Data Science and Analytics						
Semester	Fifth	Course Category Code: DSC *End Semester Exam Type: TE						
Course Code	A20DAP501	Periods / Week			Credit	Maximum Marks		
		L	T	P	C	CA M	ESE	TM
Course Name	Mini Project	4	0	0	2	50	50	100

Domains:

- Social Network Analysis
- Business Analysis
- E-Commerce Analysis
- Banking Analysis
- Digital Marketing Analysis

Sl.no	Description	Weightage		
1	Continuous Assessment Marks			
a	Review 1	Review Committee	5	10
		Guide	5	
b	Review 2	Review Committee	5	10
		Guide	5	
c	Review 3	Review Committee	15	30
		Guide	15	
Total CAM			50	
2	End Semester Marks			
a	Evaluation of Mini Project report	Internal Examiner	20	40
		External Examiner	20	
b	Outcome	Publication of Papers/ Conference Presentations/ Patents/ Prototypes etc.	10	10
Total ESM			50	
Total Marks			100	

R.D. Mithal

Department	Computational Studies			Programme: B.Sc DATA SCIENCE AND ANALYTICS						
Semester	Fifth			Course Category Code: SEC		*End Semester Exam Type: TE				
Course Code	A20DAS505			Periods / Week			Credit	Maximum Marks		
				L	T	P	C	CAM	ESE	TM
Course Name	R Programming Lab			0	0	4	2	25	75	100
Prerequisite	Basic knowledge in R Language.									
Course Objectives	<ul style="list-style-type: none"> To understand the Data Science. To understand the basic of R language. To implement statistics techniques. To implement function and Regression. To apply Nonlinear Models and Non-Linear with R. 									
Course Outcome	<i>After the completion of this course, the students will be able to:</i>							BT Mapping (Highest Level)		
	CO1	Recognize various disciplines that contribute to a successful data science effort.						K2		
	CO2	Understand the processes of data science						K3		
	CO3	Be aware of the challenges that arise in data sciences.						K3		
	CO4	Develop and appreciate various techniques for data modelling and mining.						K4		
	CO5	Be cognizant of ethical issues in many data science tasks.						K4		
UNIT-I	GETTING STARTED WITH R							Periods: 06		
Installing R - The R environment - R packages - Basics of R - Data Structures - Reading data into R - Graphics in R.								CO1		
UNIT-II	FUNCTIONS AND STATEMENTS							Periods: 06		
Writing R functions - Control Statements (if and else, switch, if else, compound tests) - Loops in R (for, while, controlling loops) - Applications using the functions and loops.								CO2		
UNIT-III	DATA MANIPULATION AND ANALYSIS							Periods: 06		
NGroup manipulation - Data Reshaping - Manipulating Strings - Basic Statistics using R (Summaries, Correlation, t-tests, ANOVA)								CO3		
UNIT-IV	LINEAR MODELS USING R							Periods: 06		
Linear Models - Simple and Multiple regression, GLM - Logit Regression, Model diagnostics - Residuals, Cross validation, Boot strapping.								CO4		
UNIT-V	IN NON-LINEAR MODELS, TIME SERIES AND CLUSTERING USING R FERENTIAL STATICS WITH R							Periods: 06		
Nonlinear Models - Non-Linear least square, Splines, Generalised Additive Models, Decision trees, Random forests. Time Series - Autoregressive moving average, VAR, GARCH. Clustering - K means, PAM and Hierarchical Clustering.								CO5		
Lecture Periods: 30			Tutorial Periods: -		Practical Periods: -			Total Periods: 30		
Text Books										
<ol style="list-style-type: none"> W. N. Venables, D. M. Smith and the R Core Team, "An Introduction to R", 2013. Tony Ojeda, Sean Patrick Murphy, Benjamin Bengfort, Abhijit Dasgupta, "Practical Data Science Cookbook", Packt Publishing Ltd., 2014. Nathan Yau, "Visualize This: The FlowingData Guide to Design, Visualization, and Statistics", Wiley, 2011. 										
Reference Books										
<ol style="list-style-type: none"> Nina Zumel, John Mount, "Practical Data Science with R", Manning Publications, 2014. Jure Leskovec, Anand Rajaraman, Jeffrey D. Ullman, "Mining of Massive Datasets", Cambridge 										

University Press, 2014.

6. Mark Gardener, "Beginning R - The Statistical Programming Language", John Wiley & Sons, Inc., 2012.

Web References

- 16. <https://www.rdocumentation.org/>
- 17. <https://www.geeksforgeeks.org/r-programming-for-data-science/>
- 18. https://onlinecourses.nptel.ac.in/noc21_cs23/preview
- 19. <https://archive.nptel.ac.in/courses/111/104/111104147/>

* TE – Theory Exam, LE – Lab Exam

COs/POs/PSOs Mapping

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

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

Assessment Pattern as per Bloom’s Taxonomy

Test / Bloom’s Category*	Remembering (K1) %	Understanding (K2) %	Applying (K3) %	Analyzing (K4) %	Evaluating (K5) %	Creating (K6) %	Total %
CAT1	10	20	70	-	-	-	100
CAT2	10	20	70	-	-	-	100
ESE	10	30	60	-	-	-	100

* ±3% may be varied

Evaluation Method

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

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

57

R.D. McWhorter

57

R.D. McWhorter