

M.Tech (Data Science) REGULATION 2019

Vision

• To be a Department of Excellence in Information Technology Education, Research and Development.

Mission

• To train the students to become highly knowledgeable in the field of Information Technology.

• To promote continuous learning and research in core and emerging areas.

• To develop globally competent students with strong foundations, who will be able to adapt to changing technologies.

PROGRAMME OUTCOMES (POs)

PO1: Graduates should be able to learn how to interpret data, extracts meaningful information, and assesses findings.

PO2: Graduates should capable of demonstrating and developing a design of mastery over the key technologies in data science and business analytics such as structured/unstructured data mining, machine learning, visualization techniques, predictive modeling and statistics.

PO3: Graduates should be capable of applying ethical principles and responsibilities during Professional practice.

PO4: Graduates should be able to function effectively as a team member and to write/ present a substantial technical report / document.

PO5: Graduates should independently carry out research / investigation and development work to solve industry and organization-specific problems and challenges using advanced analytics and computational methods.

PO6: Graduates should be able to engage in independent and life-long learning in the broadest context of technological change.

M.Tech (Data Science) CHOICE BASED CREDIT SYSTEM

Now a day's industries, such as social media, healthcare, insurance, e-commerce, transport, government, banking, telecommunications, etc., that are producing massive amounts of data, the so-called "BIG DATA", at a remarkable scale. This has led to converting unprocessed data to useful information to make decisions. In response, the **Department of Information Technology at Rajalakshmi Engineering College** offers a Master's program on Data Science.

M.Tech. Data Science is a two year postgraduate course. The course is designed to meet the needs of the IT professionals to grasp the industrial requirements and create data scientists within short time. Data Science program is to create professional who can mine and interpret data and making sense for industrial applications.

Eligibility

• B.E./B.Tech. or equivalent degree with minimum 50% marks or equivalent CGPA in CSE,IT and ECE

S. No	COURSE CODE	COURSE TITLE	CATEG ORY	CONTAC T	L	Т	Р	С
THE			0	PERIODS				
1.	MH19107	Mathematics for data Science	BS	4	3	1	0	4
2.	PG19101	Research Methodology and IPR	МС	3	3	0	0	3
3.	DS19101	Programming for Data Science	PC	3	3	0	0	3
4.	DS19102	Advanced Data Structures and Algorithmic methods	PC	3	3	0	0	3
5.	DS19103	Data Analysis and Visualization	PC	3	3	0	0	3
6.	DS19104	Scalable Systems for Data Science	PC	3	3	0	0	3
7.	AC19101	English for Research Paper Writing	HS	3	3	0	0	0
	PRACTICALS							
7.	DS19111	Advanced Data Structures and Algorithms Laboratory	PC	4	0	0	4	2
8.	DS19112	Data Analysis using Python Laboratory	PC	4	0	0	4	2
Total				30	21	1	8	23

SEMESTER I

Curriculum

SEMESTER I	Ι
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S. No	COURSE CODE	COURSE TITLE	CATEG ORY	CONTACT PERIODS	L	Т	Р	С
THE	ORY							
1.	DS19201	Big Data Analytics and Tools	PC	3	3	0	0	3
2.	DS19202	Cloud Computing Technologies	PC	3	3	0	0	3
3.	DS19203	Machine Learning Techniques	PC	3	3	0	0	3
4.	DS19204	Computer Vision	PC	3	3	0	0	3
5.	DS19P2X	Professional Elective I	PE	3	3	0	0	3
6.	DS19P2X	Professional Elective II	PE	3	3	0	0	3
7.	AC19201	Constitution of India	MC	3	3	0	0	0
PRA	CTICALS			-				
7.	DS19211	Big Data Laboratory	PC	4	0	0	4	2
8.	DS19212	Cloud Computing Technologies Laboratory	PC	4	0	0	4	2
Tota	l			29	21	0	8	22

SEMESTER III

S. No	COURSE CODE	COURSE TITLE	CATEGORY	CONTACT PERIODS	L	Т	Р	С
THE	ORY							
1.	DS19301	Security for Data Science	PC	3	3	0	0	3
2.	DS19P3X	Professional Elective III	PE	3	3	0	0	3
3.		Open Elective I	OE	3	3	0	0	3
PRA	CTICALS							
4.	DS19311	Project Work (Phase I)	EEC	12	0	0	12	6
Tota	Total			21	9	0	12	15

SEMESTER IV

S. No	COURSE CODE	COURSE TITLE	CATEGORY	CONTACT PERIODS	L	Т	Р	С
PRA	PRACTICALS							
1.	DS19411	Project Work (Phase II)	EEC	24	0	0	24	12
Total	Total			24	0	0	24	12

Total Credits: 72

LIST OF ELECTIVES

SEMESTER II-ELECTIVE I

S. No	COURSE CODE	COURSE TITLE	CATEGORY	CONTACT PERIODS	L	Т	Р	С
1.	DS19P21	Advanced Database Technology	PE	3	3	0	0	3
2.	DS19P22	Video Analytics	PE	3	3	0	0	3
3.	DS19P23	Deep Learning	PE	3	3	0	0	3
4.	DS19P24	FiniteElements:TheoryandAlgorithms	PE	3	3	0	0	3

SEMESTER II-ELECTIVE II

S. No	COURSE CODE	COURSE TITLE	CATEGORY	CONTACT PERIODS	L	Т	Р	С
1.	DS19P25	Information Retrieval Techniques	PE	3	3	0	0	3
2.	DS19P26	Natural Language Processing	PE	3	3	0	0	3
3.	DS19P27	Social Network Analysis	PE	3	3	0	0	3
4.	DS19P28	Recommendation Systems	PE	3	3	0	0	3

SEMESTER III-ELECTIVE III

S. No	COURSE CODE	COURSE TITLE	CATEGORY	CONTACT PERIODS	L	Т	Р	С
1.	DS19P31	Health Care Analytics	PE	3	3	0	0	3
2.	DS19P32	Financial and Risk Analytics	PE	3	3	0	0	3
3.	DS19P33	Marketing and Retail Analytics	PE	3	3	0	0	3
4.	DS19P34	Supply Chain and Logistics Analytics	PE	3	3	0	0	3

OPEN ELECTIVE COURSES

S. No	COURSE CODE	COURSE TITLE	CATEGORY	CONTACT PERIODS	L	Т	Р	С
1.	CP19O31	Business Analytics	OE	3	3	0	0	3
2.	ED19032	Operation Research	OE	3	2	1	0	3
3.	PG19031	Cost Management of Engineering Projects	OE	3	3	0	0	3
4.	PG19032	Waste To Energy	OE	3	3	0	0	3
5.	ED19031	Industrial Safety	OE	3	3	0	0	3
6.	ED19033	Composite Materials	OE	3	3	0	0	3

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
MH19107	MATHEMATICS FOR DATA SCIENCE	BS	3	1	0	4

Ob	jectives:						
•	To understand the concept of vector spaces and orthogonalityfor solving matrix decomposition problems.						
•	To develop the concept of correlation and regression and apply in real life problems.						
•	• To develop the skills on decision making using the concepts in estimation theory.						

UNIT-I	VECTOR SPACE	12
Vector space	ces - Inner Products and Norms - Orthogonality - Eigen decomposition of Matrices: Sim	ilarity
Transformat	tion - Orthogonal Transformation of Symmetric matrices.	
UNIT-II	MATRIX THEORY	12
The Choles	ky decomposition - generalized eigen vectors, canonical basis - QR factorization: Gram -Schmidt pro	ocess -
least squares	s method - singular value decomposition.	
UNIT-III	TWO DIMENSIONAL RANDOM VARIABLES	12
Joint distrib	utions - marginal and conditional distributions - Transformation of two dimensional random varia	bles –
regression c	urve – correlation.	
UNIT-IV	ESTIMATION THEORY	12
Unbiased E	stimators - method of moments - maximum likelihood estimation - curve fitting by principle o	f least
squares - re	gression lines.	
UNIT-V	MULTIVARIATE ANALYSIS	12
D 1	ctors and matrices - mean vectors and covariance matrices –multivariatenormal density and its prope	erties -
Random veo	tors and matrices mean vectors and covariance matrices matrixanatenormal density and its prope	_
	principal components- principle components from standardized variables.	_

	Course Outcomes: On completion of the course, students will be able to		
•	Use the concept of vector spaces and inner products.		
•	Decompose matrices and obtain canonical bases.		
•	Apply the concept of correlation and regression in real life situation.		
•	Apply the concept of estimation theory and curve fitting for forecasting.		
•	Identify and analyze the principle components of different processes.		

Ref	ference Books(s) / Web links:
1	Richard Bronson, "Matrix Operation", Schaum's outline series, 2nd Edition, McGraw
1	Hill, 2011.
2	Veerarajan T, Probability, statistics and random process with queueing theory and queueing networks, 4th
4	edition, McGraw - Hill Publishing Company Limited
2	Richard A. Johnson and Dean W. Wichern, "Applied Multivariate Statistical Analysis", Pearson Education, Asia,
3	Fifth Edition.
4	Gupta S.C. and Kapoor V.K."Fundamentals of Mathematical Statistics", Sultan and Sons.
5	Jay L. Devore, "Probability and Statistics for Engineering and the Sciences", Thomson and Duxbury.
6	Richard Johnson. "Miller & Freund's Probability and Statistics for Engineer", Prentice - Hall, Seventh Edition,
0	2007.
7	Dallas E Johnson, "Applied Multivariate Methods for Data Analysis", Thomson and Duxburypress.

РО	PO1	PO2	DO3	D O4	PO5	PO6
СО	FOI	FO2	PO3	PO4	POS	r00
C01	2	2	-	-	3	3
C02	1	2	-	-	3	2
C03	2	1	-	-	3	3
C04	2	2	-	-	3	3
C05	2	1	-	-	3	2
Average	1.8	1.6	0	0	3	2.6

Correlation levels 1, 2 or 3 are as defined below:

1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High)

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	C
PG19101	Research Methodology And IPR	MC	3	0	0	3
Objectives:		·				
• IPR is to	ate the importance of research methodology and Intellectual Property Rights make the students aware of their rights for the protection of their invention do gistration of patents in our country and foreign countries of invention, de	ne in their proj	ect	woi	rk.	
	o get knowledge of patents, copy right, trademarks and designs.	orgino uno uno				. ,
UNIT-I	Research Methodology: Meaning of research problem, Sources of research Characteristics of a good research problem, Errors in selecting a research p bjectives of research problem. Approaches of investigation of solutions fo lata collection, analysis, interpretation, Necessary instrumentations.	problem, Scop	e ai	nd	9	
]	Review Of Literature And Technical Writing: Effective literature studies a Plagiarism, Research ethics, Effective technical writing, how to write report, Research Proposal, Format of research proposal, a presentation and asses ommittee.	Paper Develop	oing	a	9	
1	ntellectual Property Rights: Nature of Intellectual Property: Patents, E Copyright, copyright registration in India Process of Patenting and Develop esearch, innovation, patenting, development. International Scenario: Internati ntellectual Property. Procedure for grants of patents, Patenting under Patent C	ment: technolo onal cooperation	ogic on c	al on	9	
	Patent Rights And Recent Developments In IPR: Patent Rights: Scope of Pate				9	
]	nd transfer of technology. Patent information and databases. Geographica Developments in IPR: Administration of Patent System. New developme Biological Systems, Computer Software etc. Traditional knowledge Case Stud	nts in IPR; IF	PR			
UNIT-V	ndustrial Designs And Geographical Indications : Industrial designs and Registrations of designs, conditions and procedures of industrial design Registration, International convention of design- types and functions. Semic ircuits and layout design Act- Geographical indications-potential benefindications.	IC Layout de ns- Cancellation conductor Integ	esig on grate	of ed	9	

Co	urse Outcomes:			
•	Student can understand the research problem formulation and analyze research related information.			
•	Understanding that when IPR would take such important place in growth of individuals & nation.			
•	Understand the importance of copyright and industrial designs.			
•	Understand that IPR protection provides an incentive to inventors for further research work and investment in R & D, which leads to creation of new and better products, and in turn brings about, economic growth and social benefits.			
•	The students once they complete their academic projects, they get awareness of acquiring the patent and copyright for their innovative works.			

Tex	xt Book (s):
1	Neeraj Pandey and Khushdeep Dharni, Intellectual Property Rights, First edition, PHI learning Pvt. Ltd., Delhi, 2014.
2	Uma Sekaran and Roger Bougie, Research methods for Business, 5th Edition, Wiley India, New Delhi, 2012.
3	Stuart Melville and Wayne Goddard, "Research methodology: an introduction for science & engineering students" ,2 nd edition, Juta Academic, 2001.
4	Ramakrishna B & Anilkumar H S, Fundamentals of Intellectual Property Rights, Ist edition, Notion Press, 2017.

Reference Books(s) / Web links:

William G Zikmund, Barry J Babin, Jon C.Carr, Atanu Adhikari,Mitch Griffin, Business
 Research methods, A South Asian Perspective, 8th Edition, Cengage Learning, New Delhi, 2012.

CO-PO MAPPING

PO						
со	PO1	PO2	PO3	PO4	PO5	PO6
C01	2	1	2	2	3	2
C02	2	2	2	2	2	2
C03	1	1	1	1	2	2
C04	1	2	2	2	3	2
C05	1	1	1	2	2	2
Average	2	1	2	2	2	2

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19101	PROGRAMMING FOR DATA SCIENCE	PC	3	0	0	3

01	
()hi	ectives:
00	ccuves.

• To understand the basics of Data scien
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• To learn the basics of python programming.

• To study the data structures provided by numpy library for arrays and Visualization using Python.

• To learn the basics of R programming.

• To learn and implement Data Science algorithms using R.

UNI	ΤI	INTRODUCTION TO DATA SCIENCE	9	9
Data	Science	e - Evolution of data – Best Practices for Big data Analytics – Big data characteristics – Vali	lating – '	The
Pron	notion o	f the Value of Big Data – Big Data Use Cases- Characteristics of Big Data Application-	Statistics	for
		Data science life cycle – Analysis and reporting –Modern Data Analytics Tools		
UNI	TII	INTRODUCTION TO PYTHON PROGRAMMING	9	9
Basi	cs of P	ython-Starting with python interpreter-Control flow statements-Functions-List,tuple,Diction	ary-Lam	ıbda
with	mapper	reduce and filter-List and dictionary comprehension- Files- Modules -Packages - Regular e	xpression	ons -
Wor		h classes and Objects		
UNI	TIII	DATA SCIENCE USING PYTHON	9	9
Data	Scienc	e Packages-Numpy, scipy, pandas-Building models and evalution with Scikit-Data Loading.	Storage	and
		Data Wrangling: Clean, Transform, Merge, Reshape-Plotting and Visualization - Data Agg		
		ations - Time Series - The Jupyter and PyDev development environments-Neural Network E		
		in Python - Statistical Methods for Evaluation using R - Visualization using Python - Building	models	and
evalu	ution wi	th Scikit		
UNI	T IV	INTRODUCTION TO R	9	9
			-	-
Over	rview- I	DataTypes, Variables, Operators, Decision Making, Loops, Functions, Strings, Vector, Li	st, Matri	ces,
Over Arra	rview- I ys, Fact		st, Matri	ces,
Over Arra	rview- I ys, Fact oratory	DataTypes, Variables, Operators, Decision Making, Loops, Functions, Strings, Vector, Licors, Data Frames, Packages-Data Visualization – Data Interfaces- Reading and writing of Data Analysis using R-Statistical Methods for Evaluation using R	st, Matrie CSV fi	ces,
Over Arra Expl	rview- I ys, Fact oratory T V	DataTypes, Variables, Operators, Decision Making, Loops, Functions, Strings, Vector, Licors, Data Frames, Packages-Data Visualization – Data Interfaces- Reading and writing of Data Analysis using R-Statistical Methods for Evaluation using R DATA SCIENCE USING R	st, Matrie CSV fi	ices, iles- 9
Over Arra Expl UNI Asso	rview- I ys, Fact oratory T V ociation	DataTypes, Variables, Operators, Decision Making, Loops, Functions, Strings, Vector, Licors, Data Frames, Packages-Data Visualization – Data Interfaces- Reading and writing of Data Analysis using R-Statistical Methods for Evaluation using R DATA SCIENCE USING R –Classification – Clustering – Time series Analysis-Text Analysis-–Prediction Algo	st, Matrie CSV fi	ices, iles- 9
Over Arra Expl UNI Asso	rview- I ys, Fact oratory T V ociation	 DataTypes, Variables, Operators, Decision Making, Loops, Functions, Strings, Vector, Litors, Data Frames, Packages-Data Visualization – Data Interfaces- Reading and writing of Data Analysis using R-Statistical Methods for Evaluation using R DATA SCIENCE USING R Classification – Clustering – Time series Analysis-Text Analysis-—Prediction Algo Video Analytics – Data base analytics 	st, Matric CSV fi g rithm-Im	ices, iles- 9 nage
Over Arra Expl UNI Asso Anal	rview- I ys, Fact oratory T V ociation lytics – V	DataTypes, Variables, Operators, Decision Making, Loops, Functions, Strings, Vector, Li tors, Data Frames, Packages-Data Visualization – Data Interfaces- Reading and writing of Data Analysis using R-Statistical Methods for Evaluation using R DATA SCIENCE USING R –Classification – Clustering – Time series Analysis-Text Analysis-–Prediction Algo Video Analytics – Data base analytics Total Contact Hours	st, Matric CSV fi g rithm-Im	ices, iles- 9
Over Arra Expl UNI Asso Anal	rview- I ys, Fact oratory T V ociation lytics – V rse Out	DataTypes, Variables, Operators, Decision Making, Loops, Functions, Strings, Vector, Li tors, Data Frames, Packages-Data Visualization – Data Interfaces- Reading and writing of Data Analysis using R-Statistical Methods for Evaluation using R DATA SCIENCE USING R –Classification – Clustering – Time series Analysis-Text Analysis-–Prediction Algo Video Analytics – Data base analytics Total Contact Hours comes:	st, Matric CSV fi g rithm-Im	ices, iles- 9 nage
Over Arra Expl UNI Asso Anal	rview- I ys, Fact oratory T V ociation lytics – ' rse Out complet	DataTypes, Variables, Operators, Decision Making, Loops, Functions, Strings, Vector, Li tors, Data Frames, Packages-Data Visualization – Data Interfaces- Reading and writing of Data Analysis using R-Statistical Methods for Evaluation using R DATA SCIENCE USING R –Classification – Clustering – Time series Analysis-Text Analysis-–Prediction Algo Video Analytics – Data base analytics Total Contact Hours comes: ion of course students will be able to	st, Matric CSV fi g rithm-Im	ices, iles- 9 nage
Over Arra Expl UNI Asso Anal Cou On c	rview- I ys, Fact oratory T V ociation lytics – ' rse Out complet has exp	DataTypes, Variables, Operators, Decision Making, Loops, Functions, Strings, Vector, Li tors, Data Frames, Packages-Data Visualization – Data Interfaces- Reading and writing of Data Analysis using R-Statistical Methods for Evaluation using R DATA SCIENCE USING R –Classification – Clustering – Time series Analysis-Text Analysis—Prediction Algo Video Analytics – Data base analytics Total Contact Hours comes: ion of course students will be able to erience in implementation/modification of methods involved in Data Science	st, Matric CSV fi g rithm-Im	ices, iles- 9 nage
Over Arra Expl UNI Asso Anal Cou On c	rview- I ys, Fact oratory T V Deciation lytics – V rse Out complet has exp Apply p	DataTypes, Variables, Operators, Decision Making, Loops, Functions, Strings, Vector, Li tors, Data Frames, Packages-Data Visualization – Data Interfaces- Reading and writing of Data Analysis using R-Statistical Methods for Evaluation using R DATA SCIENCE USING R –Classification – Clustering – Time series Analysis-Text Analysis-–Prediction Algo Video Analytics – Data base analytics Total Contact Hours comes: ion of course students will be able to erience in implementation/modification of methods involved in Data Science programming knowledge of python	st, Matric CSV fi g rithm-Im	ices, iles- 9 nage
Over Arra Expl UNI Asso Anal Cou On c • •	rview- I ys, Fact oratory T V ociation lytics – V rse Out complet has exp Apply p manage	DataTypes, Variables, Operators, Decision Making, Loops, Functions, Strings, Vector, Li tors, Data Frames, Packages-Data Visualization – Data Interfaces- Reading and writing of Data Analysis using R-Statistical Methods for Evaluation using R DATA SCIENCE USING R –Classification – Clustering – Time series Analysis-Text Analysis-–Prediction Algo Video Analytics – Data base analytics Total Contact Hours comes: ion of course students will be able to erience in implementation/modification of methods involved in Data Science programming knowledge of python e data , analysis and problem solving	st, Matric CSV fi g rithm-Im	ices, iles- 9 nage
Over Arra Expl UNI Asso Anal	rview- I ys, Fact oratory T V ociation ytics – V rse Out complet has exp Apply p manage Gain k	DataTypes, Variables, Operators, Decision Making, Loops, Functions, Strings, Vector, Li tors, Data Frames, Packages-Data Visualization – Data Interfaces- Reading and writing of Data Analysis using R-Statistical Methods for Evaluation using R DATA SCIENCE USING R –Classification – Clustering – Time series Analysis-Text Analysis-–Prediction Algo Video Analytics – Data base analytics Total Contact Hours comes: ion of course students will be able to erience in implementation/modification of methods involved in Data Science programming knowledge of python	st, Matric CSV fi g rithm-Im	ices, iles- 9 nage

Reference Books(s) / Web links:

1	Wes McKinney, "Python for Data Analysis", O'Reilly Media.2012
2	Sebastian Raschka, "Python Machine Learning", Packpub.com, 2015
3	https://www.datacamp.com/courses/statistical-thinking-in-python-part-1

PO	DOI		DOD			DO
со	PO1	PO2	PO3	PO4	PO5	PO6
C01	3	3	1	1	2	2
C02	3	3	1	1	3	1
C03	3	3	1	1	3	1
C04	3	3	1	1	3	2
C05	3	3	1	1	3	2
Average	3	3	1	1	2.8	1.6

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19102	ADVANCED DATA STRUCTURES AND ALGORITHMIC	PC	3	0	0	3
	METHODS					

Ob	Objectives:					
•	derstand the principles of iterative and recursive algorithms.					
•	rn the graph search algorithms.					
٠	dy network flow and linear programming problems.					
٠	rn the hill climbing and dynamic programming design techniques.					
•	velop recursive backtracking algorithms.					

UNIT I **ITERATIVE AND RECURSIVE ALGORITHMS**

Iterative Algorithms: Measures of Progress and Loop Invariants-Paradigm Shift: Sequence of Actions versus Sequence of Assertions- Steps to Develop an Iterative Algorithm-Different Types of Iterative Algorithms--Typical Errors-Recursion-Forward versus Backward- Towers of Hanoi Checklist for Recursive Algorithms-The Stack Frame-Proving Correctness with Strong Induction Examples of Recursive Algorithms-Sorting and Selecting Algorithms-Operations on Integers Ackermann's Function- Recursion on Trees-Tree Traversals- ExamplesGeneralizingtheProblem-HeapSortandPriorityQueues-RepresentingExpressions.

OPTIMISATION ALGORITHMS UNIT II

Optimization Problems-Graph Search Algorithms-Generic Search-Breadth-First Search Dijkstra's Shortest-Weighted-Path-Depth-First Search-Recursive Depth-First Search-Linear Ordering of a Partial Order- Network Flows and Linear Programming-Hill Climbing-Primal Dual Hill Climbing Steepest Ascent Hill Climbing-Linear Programming-Recursive Backtracking-Developing Recursive Backtracking Algorithm- Pruning Branches-Satisfiability. 9

UNIT III DYNAMIC PROGRAMMING ALGORITHMS

Developing a Dynamic Programming Algorithm-Subtle Points- Question for the Little Bird Sub instances and Sub solutions -Set of Substances-Decreasing Time and Space-Number of Solutions-Code. Reductions and NP Completeness – Satisfiability - Proving NPCompleteness-3-Coloring- Bipartite Matching. Randomized Algorithms - Randomness to Hide Worst Cases Optimization Problems with a Random Structure

SHARED OBJECTS AND CONCURRENT OBJECTS UNIT IV

Shared Objects and Synchronization -Properties of Mutual Exclusion-The Mora 1- The Producer-Consumer Problem - The Readers-Writers Problem-Realities of Parallelization Parallel Programming- Principles- Mutual Exclusion-Time- Critical Sections-Thread Solutions-The Filter Lock-Fairness-Lamport's Bakery Algorithm-Bounded Timestamps-Lower Bounds on the Number of Locations-Concurrent Objects-Concurrency and Correctness Sequential Objects-Quiescent Consistency-Sequential Consistency-Linearizability-Formal Definitions- Progress Conditions- The Java Memory Model.

CONCURRENT DATA STRUCTURES UNIT V

Practice-Linked Lists-The Role of Locking-List-Based Sets-Concurrent Reasoning- Coarse Grained Synchronization-Fine-Grained Synchronization-Optimistic Synchronization- Lazy Synchronization-Non-Blocking Synchronization-Concurrent Queues and the ABA Problem Queues-A Bounded Partial Queue-An Unbounded Total Queue-An Unbounded Lock-Free Queue Memory Reclamation and the ABA Problem- Dual Data Structures- Concurrent Stacks and Elimination- An Unbounded Lock-Free Stack- Elimination-The Elimination Backoff Stack.

Total Contact Hours : 45

9

9

Cou	ourse Outcomes:					
On	completion of course students will be able to					
•	Design and apply iterative and recursive algorithms.					
•	Design and implement optimization algorithms in specific applications.					
•	Design appropriate shared objects and concurrent objects for applications.					
•	Implement and apply concurrent linked lists, stacks, and queues.					

Refe	erence Books(s) / Web links:						
1	Jeff Edmonds, "How to Think about Algorithms", Cambridge University Press, 2008.						
2	M. Herlihy and N. Shavit, "The Art of Multiprocessor Programming", MorganKaufmann, 2008.						
3	Steven S. Skiena, "The Algorithm Design Manual", Springer, 2008.						
4	Peter Brass, "Advanced Data Structures", Cambridge University Press, 2008.						
5	S. Dasgupta, C. H. Papadimitriou and U. V. Vazirani, "Algorithms", McGraw-Hill, 2008.						
6	J. Kleinberg and E. Tardos, "Algorithm Design", Pearson Education, 2006.						
7	T. H. Cormen, C. E. Leiserson, R. L. Rivest and C. Stein, "Introduction to Algorithms", PHI Learning Private						
/	Limited, 2012.						
8	Rajeev MotwaniandPrabhakarRaghavan, "Randomized Algorithms", CambridgeUniversity Press, 1995.						
9	A. V. Aho, J. E. Hopcroft, and J. D. Ullman, "The Design and Analysis of ComputerAlgorithms", Addison-						
9	Wesley, 1975.						
10	A. V. Aho, J. E. Hopcroft, and J. D. Ullman,"Data Structures and Algorithms", Pearson, 2006.						

РО	PO1	PO2	PO3	PO4	PO5	PO6
СО	FOI	FO2	ros	r04	POS	rOo
C01	3	3	1	2	3	2
C02	3	3	1	2	3	2
C03	3	3	1	2	3	2
C04	3	3	1	2	3	2
C05	3	3	1	2	3	2
Average	3	3	1	2	3	2

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Theory course) C		L	Т	Р	С
DS19103	DATA ANALYSIS AND VISUALIZATION	PC	3	0	0	3

Objectives

U U	jecuves.						
•	o parse a real-world data analysis problem into one or more computational components learned in this course,						
	To apply suitable machine learning and/or visualization techniques and analyse the results obtained to enable						
•	optimal decision making.						
•	To understand regression and classification algorithms.						
•	To learn data visualization techniques.						

INTRODUCTION TO DATA ANALYTICS UNIT I

Importance of analytics and visualization in the era of data abundance- -Review of probability, statistics and random processes-- Brief introduction to estimation theory

MACHINE LEARNING TECHNIQUES UNIT II

Introduction to machine learning, supervised and unsupervised learning, gradient descent, overfitting, regularization etc. - Clustering techniques: K-means, Gaussian mixture models and expectation-maximization, agglomerative clustering, evaluation of clustering - Rand index, mutual information based scores, Fowlkes-Mallows index etc.

UNIT III REGRESSION AND CLASSIFICATION

Regression: Linear models, ordinary least squares, ridge regression, LASSO, Gaussian Processes regression. -Supervised classification methods: K-nearest neighbor, naive Bayes, logistic regression, decision tree, support vector machine. - Sparse coding and dictionary learning, orthogonal matching pursuit.- Introduction to artificial neural networks (ANNs), deep NNs, convolutional neural network (CNN), and other recent topics.

UNIT IV DATA VISUALIZATION

Data visualization: Basic principles, categorical and continuous variables. - Exploratory graphical analysis. - Creating static graphs, animated visualizations - loops, GIFs and Videos.

UNIT V DATA VISUALIZATION TOOLS

Data visualization in Python and R, examples from Bokeh, Altair, ggPlot, ggplot2, gganimate, ImageMagick etc. -Introduction to Visualization Toolkit (VTK) for 3D computer graphics, image processing and visualization. -Visualization for deep learning. 45

Total Contact Hours

9

9

9

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Course Outcomes:

On completion of course students will be able to

٠	Parse a real-world data analysis problem i	nto one or more computational c	components learned in this course.
			· · · · · · · · · · · · · · · · · · ·

- Apply suitable machine learning and/or visualization techniques and analyze the results obtained to enable . optimal decision making.
- Understand regression and classification algorithms. •
- Apply data visualization techniques. •
- Use Data visualization tools

Reference Books(s) / Web links:

1	Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, The Elements of Statistical Learning, Springer, 2001.					
2	Christopher Bishop, Pattern Recognition and Machine Learning, Springer, 2006.					
3	David G. Stork, Peter E. Hart, and Richard O. Duda, Pattern Classification (2nd edition), Wiley, 2000.					
4	. Edward Tufte, The Visual Display of Quantitative Information (2nd edition), Graphics Press, 2001.					
5	Colin Ware, Information Visualization: Perception for Design (2nd edition), Morgan Kaufmann, 2004.					
6	Alberto Cairo, The Functional Art: An Introduction to Information Graphics and Visualization, New Riders,					
0	Pearson Education, 2013.					
7	Nathan Yau, Data Points: Visualization That Means Something, Wiley, 2013.					
8	Charles D. Hansen and Chris R. Johnson, Visualization Handbook, Academic Press, 2004.					
9	Will Schroeder, Ken Martin, and Bill Lorensen, The Visualization Toolkit: An Object-Oriented Approach to 3D					
9	Graphics, Kitware Inc. Publishers, 2004.					

PO	PO1	PO2	PO3	PO4	PO5	PO6
СО	POI	PO2	POS	r04	r05	POO
C01	3	3	-	2	3	2
C02	3	3	-	2	3	2
C03	3	3	-	2	3	2
C04	3	3	-	2	3	2
C05	3	3	-	2	3	2
Average	3	3	-	2	3	2

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Theory course)		L	Т	Р	С
DS19104	SCALABLE SYSTEMS FOR DATA SCIENCE	PC	3	0	0	3

Ob	Objectives:		
•	To understand the basics of data science.		
•	• To understand advanced analytical theory and methods.		
•	To learn the advanced analytics technology and tools		
•	To get exposed with the Hadoop distributed file system architecture		
•	To understand the Map-Reduce concepts.		

UNIT I	INTRODUCTION TO DATA SCIENCE	9				
Introduction	Introduction of Data Science, Basic Data Analytics using R, R Graphical User Interfaces- Data Import and Export,					
Attribute and	d Data Types, Descriptive Statistics, Exploratory Data Analysis, Visualization Before Analysis,	Dirty				
Data, Visual	izing a Single Variable, Examining Multiple Variables, Data Exploration Versus Presentation- Stat	istical				
Methods for	Evaluation, Hypothesis Testing, Difference of Means, Wilcoxon Rank-Sum Test, ANOVA					
UNIT II	ADVANCED ANALYTICAL THEORY AND METHODS	9				
Overview of	f Clustering, K-means, Use Cases, Overview of the Method, Perform a K-means Analysis usi	ng R-				
Classification	n, Decision Trees, Overview of a Decision Tree, Decision Tree Algorithms, Evaluating a Decision	Tree-				
Decision Tre	e in R, Bayes' Theorem, Naïve Bayes Classifier, Smoothing, Naïve Bayes in R					
UNIT III	ADVANCED ANALYTICS TECHNOLOGY AND TOOLS	9				
Analytics for	r Unstructured Data, Use Cases, MapReduce, Apache Hadoop, The Hadoop Ecosystem, Pig, Hive, H	Ibase,				
Mahouth, No	oSQL, SQL Essentials- Joins, Set Operations, Grouping Extensions, In-Database Text Analysis, Adv	anced				
SQL, Windo	w Functions, User-defined Functions and Aggregates, Ordered Aggregates, MADlib					
UNIT IV	HADOOP DISTRIBUTED FILE SYSTEM ARCHITECTURE	9				
HDFS Arch	itecture, HDFS Concepts, Blocks- NameNode, Secondary NameNode, DataNode, HDFS Feder	ation,				
HDFS High	Availability, Basic File System Operations- Data Flow, Anatomy of File Read, Anatomy of File	Write,				
Anatomy of	Anatomy of a MapReduce Job Run					
UNIT V	PROCESSING DATA WITH MAPREDUCE	9				
Getting to	Getting to know MapReduce, MapReduce Execution Pipeline, Runtime Coordination and Task Management-					
	MapReduce Application, Hadoop Word Count Implementation- Installing and Running Pig, Hbase Versus RDBMS,					
Installing and	d Running ZooKeeper					
	Total Contact Hours :	45				

Co	urse Outcomes:					
On	On completion of course students will be able to					
•	Understand the basics of data science.					
•	Apply theadvanced analytical theory and methods.					
•	Understand and use the advanced analytics technology and tools.					
•	Understand the Hadoop distributed file system architecture					
•	Apply the Map-Reduce concepts.					
Ref	Cerence Books(s) / Web links:					
1	David Dietrich, Barry Heller and Beibei Yang, "Data Science and Big Data Analytics: Discovering, Analyzing, Visualizing and Presenting Data", EMC Education Services, Reprint 2015, Wiley, ISBN: 9788126556533.					
2	Tom White, "Hadoop: The Definitive Guide", 4th Edition, 2015, O'Reilly, ISBN: 9789352130672.					
3	BirisLublinsky, Kevin T. Smith and Alexey Yakubovich, "Professional Hadoop Solutions", Reprint 2014, Wiley,					
3	ISBN 13:9788126551071					
4 Stephen Marsland, "Machine Learning – An Algorithmic Perspective", , Taylor& Francis Group						
-	2015, Chapman & Hall / CRC Press, ISBN:9781466583283.					
5	Nathan Marz, James Warren, "Big Data-Principles and best practices of scalable real-time data systems", Edition					
Э	2015, DreamTech Press, ISBN: 9789351198062.					

PO						
со	PO1	PO2	PO3	PO4	PO5	PO6
C01	3	3	1	2	3	2
C02	3	3	1	2	3	2
C03	3	3	1	2	3	2
C04	3	3	1	2	3	2
C05	3	3	1	2	3	2
Average	3	3	1	2	3	2

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name	Category	L	Т	Р	C
AC19101	English for Research Paper Writing	HS	3	0	0	0
	Common to all branches of M.E. /M.Tech / MBA – I Semester					

Ob	Objectives:	
•	Express technical ideas in writing	
•	Plan and organize the research paper	
•	Understand the structure and familiarise the mechanics of organised writing	
•	Improvise academic English and acquire research writing skills	

UNIT-I	INTRODUCTION TO RESEARCH WRITING		9		
Research -	Research - Types of Research - Selecting the Primary resources - Categorizing secondary sources - Discovering a				
researchable	area and topic - Need Analysis - Research Question- Focussing o	n the Research Problem- De	veloping		
Research De	sign - Framing the Hypothesis - Identifying the Scope of the Resear	ch - Writing – General and A	cademic		
Writing.		-			
UNIT-II	LANGUAGE OF WRITING		9		
Active read	ng - text mining - use of academic words - jargons - ambiguities	- use of expression - use o	f tense -		
proper voice	s - third person narration - phraseology - use of foreign words - use	of quotes - interpreting quot	es.		
UNIT-III	THE FORMAT OF WRITING		9		
Types of Jor	urnals - different formats and styles - IEEE format - Structure – Mar	gins - Text Formatting - Head	ling and		
Title - Runn	ing Head with Page Numbers - Tables and illustrations - Paper and I	Printing - Paragraphs - Highli	ghting –		
Quotation -	Footnotes				
UNIT-IV	ORGANISING A RESEARCH PAPER		9		
Title- Abstr	act – Introduction – Literature review - Methodology - Results –Di	scussion -Conclusion - Appe	ndices -		
Summarisin	g - Citation and Bibliography.				
UNIT-V	UNIT-V PUBLISHING PAPER 9				
Finding the	Finding the Prospective publication or Journal - analysing the credits - Reviewing - Revising – Plagiarism Check -				
Proof readin	Proof reading - Preparing the Manuscript- Submitting - Resubmitting - Follow up – Publishing.				
		Total Contact Hours	: 45		
			: 45		

Cou	Course Outcomes:				
On	On completion of course students will be able to				
•	Understand the basic structure of research work				
•	Apply proper use of language in writing paper				
•	Comprehend different formats of journal paper				
•	Learn the process of writing a research paper				
•	Know the process of publishing journal paper				

Reference Books / Web links:

create books / Web miks.			
Adrian Wallwork: "English for Writing Research Papers", Springer Science Business Media, Second Edition,			
LLC 2011			
Stephen Howe and Kristina Henrikssion: "Phrasebook for Writing Papers and Research in English", The Whole			
World Company Press, Cambridge, Fourth edition 2007			
The Modern Language Association of America: "MLA Handbook for Writers of Research Papers" 8th Edition,			
The Modern Language Association of America, 2016			
Rowena Murray: The Handbook of Academic Writing: A Fresh Approach, Sarah Moore			
Open University Press, 2006			
Stephen Bailey: Academic Writing: A Practical Guide for Students Routledge Falmer: 2003			
Joseph M. Moxley: Publish, Don't Perish: The Scholar's Guide to Academic Writing and Publishing, Praeger			

РО	PO1	PO2	PO3	PO4	PO5	PO6
СО		102	105	104	105	100
C01	-	1	-	-	3	3
C02	-	-	-	-	3	1
C03	-	2	-	-	3	3
C04	-	1	-	-	2	3
C05	-	1	-	-	3	2
Average	0	1.25	0	0	2.8	2.4

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Laboratory Course)	Category	L	Т	Р	С
DS19111	ADVANCED DATA STRUCTURES AND ALGORITHMS	PC	0	0	4	2
	LABORATORY					

Obj	Objectives:				
•	To learn to implement iterative and recursive algorithms.				
•	To learn to design and implement algorithms using hill climbing and dynamic programming techniques.				
•	To learn to implement shared and concurrent objects.				
•	To learn to implement concurrent data structures.				

List	t of Experiments					
	Each student has to work individually on assigned lab exercises. It is recommended that all implementations are carried out in Java. If C or C++ has to be used, then the threads library will be required for concurrency. Exercises should be designed to cover the following topics					
1.	Implementation of graph search algorithms.					
2.	Implementation and application of network flow and linear programming problems.					
3.	Implementation of algorithms using the hill climbing and dynamic programming design techniques.					
4.	Implementation of recursive backtracking algorithms.					
5.	Implementation of randomized algorithms.					
6.	Implementation of various locking and synchronization mechanisms for concurrent linked lists, concurrent queues, and concurrent stacks.					
7.	Developing applications involving concurrency					
	Total Contact Hours : 60					

Cou	urse Outcomes:
•	Design and apply iterative and recursive algorithms.
•	Design and implement algorithms using the hill climbing and dynamic programmingand recursive backtracking
-	techniques.
•	Design and implement optimisation algorithms for specific applications.
•	Design and implement randomized algorithms.
•	Implement and apply concurrent linked lists, stacks, and queues.

РО	DO1	DOA	DOA	DO 4	D 05	DOC
со	PO1	PO2	РОЗ	PO4	PO5	PO6
C01	3	3	1	2	3	2
C02	3	3	1	2	3	2
C03	3	3	1	2	3	2
C04	3	3	1	2	3	2
C05	3	3	1	2	3	2
Average	3	3	1	2	3	2

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Laboratory Course)	Category	L	Т	Р	С
DS19112	DATA ANALYSIS USING PYTHON LABORATORY	PC	0	0	4	2

Obj	jectives:							
•	To Introduce Basic Tools In Python.							
•	To Analyse the Data Using Statistical Methods							
•	To develop the Skills in using Recent Machine Learning Tools for Solving practical problems							
•	To develop the skills in applying appropriate supervised, semi-supervised or							
List	t of Experiments							
	Analytics Tools And Techniques With Python							
1.	Python: Data Manipulation, Data Exploration & Data Preparation,							
	Modelling With Python							
2.	Statistical Analysis							
	Advanced Analytics And Machine Learning							
	Linear Regression							
3.	Logistic Regression							
5.	Cluster Analysis							
	Decision Trees And Resampling Techniques							
	Ensemble Models, Boosting And Random Forest							
	Complex Techniques In Statistics And ML							
4.	Time Series							
	PCA And SVM							
	Text Mining And NLP							
5.	Data Visualization Tools							
5.	Python tools to visualize the result							
6.	Deep Learning Techniques And Tools							
7.	Real Time Case Studies Using ML in Python							
	Total Contact Hours : 60							

Cou	ourse Outcomes:					
•	Use the python tools for analysis.					
•	Apply statistical methods to understand the data					
٠	Perform data analysis with machine learning methods					
•	Perform graphical data analysis					
•	ild Analytical Models using Machine Learning and Deep learning methods.					
	Software :					
	Python Tools: NumPy, Pandas, Matplot					
	Environment Framework: Anaconda-Jupyter Notebook					
	DL frameworks: Keras, Tensor flow					

PO CO	PO1	PO2	РОЗ	PO4	PO5	PO6
C01	3	3	-	2	2	2
C02	3	3	-	2	2	2
C03	3	3	-	2	3	2
C04	3	3	-	2	3	2
C05	3	3	-	2	3	2
Average	3	3	0	2	3	2

Correlation levels 1, 2 or 3 are as defined below:

1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High)

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19201	BIG DATA ANALYTICS AND TOOLS	PC	3	0	0	3

Objectives:

• Understand big data for business intelligence

• Learn business case studies for big data analytics

• Understand nosql big data management

• Perform map-reduce analytics using Hadoop and related tools

UNIT I UNDERSTANDING BIG DATA

What is big data – why big data – convergence of key trends – unstructured data – industry examples of big data – web analytics – big data and marketing – fraud and big data – risk and big data – credit risk management – big data and algorithmic trading – big data and healthcare – big data in medicine – advertising and big data – big data technologies – introduction to Hadoop – open source technologies – cloud and big data – mobile business intelligence – Crowd sourcing analytics – inter and trans firewall analytics

UNIT II NOSQL DATA MANAGEMENT

Introduction to NoSQL – aggregate data models – aggregates – key-value and document data models – relationships – graph databases – schemaless databases – materialized views – distribution models – sharding – master-slave replication – peer-peer replication – sharding and replication – consistency – relaxing consistency – version stamps – map reduce – partitioning and combining – composing map-reduce calculations

UNIT III BASICS OF HADOOP

Data format – analyzing data with Hadoop – scaling out – Hadoop streaming – Hadoop pipes –design of Hadoop distributed file system (HDFS) – HDFS concepts – Java interface – data flow –Hadoop I/O – data integrity – compression – serialization – Avro – file-based data structures

UNIT IV	MAPRED	OUCE API	PLICATIONS	

MapReduce workflows – unit tests with MRUnit – test data and local tests – anatomy of Map Reduce job run – classic Map-reduce – YARN – failures in classic Map-reduce and YARN –job scheduling – shuffle and sort – task execution – MapReduce types – input formats – output formats

UNIT V HADOOP RELATED TOOLS

Hbase – data model and implementations – Hbase clients – Hbase examples – praxis.Cassandra– cassandra data model – cassandra examples – cassandra clients – Hadoop integration- Pig – Grunt – pig data model – Pig Latin – developing and testing Pig Latin scripts-Hive – data types and file formats – HiveQL data definition – HiveQL data manipulation – HiveQL queries.

Total Contact Hours:45

9

0

Co	urse Outcomes:
On	completion of course students will be able to
•	Describe big data and use cases from selected business domains
•	Explain NoSQL big data management
•	Install, configure, and run Hadoop and HDFS
•	Perform map-reduce analytics using Hadoop
•	Use Hadoop related tools such as HBase, Cassandra, Pig, and Hive for big dataanalytics
Ref	ference Books(s) / Web links:
1	Michael Minelli, Michelle Chambers, and AmbigaDhiraj, "Big Data, Big Analytics: Emerging Business
I	Intelligence and Analytic Trends for Today's Businesses", Wiley, 2013.
2	P. J. Sadalage and M. Fowler, "NoSQL Distilled: A Brief Guide to the Emerging World of Polyglot Persistence",
2	Addison-Wesley Professional, 2012.
3	Tom White, "Hadoop: The Definitive Guide", Third Edition, O'Reilley, 2012.
4	Eric Sammer, "Hadoop Operations", O'Reilley, 2012.
5	E. Capriolo, D. Wampler, and J. Rutherglen, "Programming Hive", O'Reilley, 2012
6	Lars George, "HBase: The Definitive Guide", O'Reilley, 2011.
7	Eben Hewitt, "Cassandra: The Definitive Guide", O'Reilley, 2010.
-	

8 Alan Gates, "Programming Pig", O'Reilley, 2011.

РО	DO1	DO2	DOJ	D O4	DO 5	DO(
со	PO1	PO2	PO3	PO4	PO5	PO6
C01	1	1	1	1	2	2
C02	2	2	0	0	2	2
C03	2	2	0	0	1	1
C04	3	3	1	1	2	2
C05	3	3	1	1	3	3
Average	2.2	2.2	0.6	0.6	2	2

Correlation levels 1, 2 or 3 are as defined below:

1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High)

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19202	CLOUD COMPUTING TECHNOLOGIES	PC	3	0	0	3

Ob	Objectives:		
•	To introduce the broad perceptive of cloud architecture and model		
٠	To understand the concept of Virtualization		
٠	To be familiar with the lead players in cloud		
٠	To understand the features of cloud simulator		
•	To apply different cloud programming model as per need		

CLOUD ARCHITECTURE AND MODEL UNIT I

Technologies for Network-Based System - System Models for Distributed and Cloud Computing - NIST Cloud Computing Reference Architecture. Cloud Models- Characteristics - Cloud Services - Cloud models (IaaS, PaaS, and SaaS) - Public vs Private Cloud - Cloud Solutions - Cloud ecosystem - Service management - Computing on demand.

UNIT II VIRTUALIZATION

Basics of Virtualization - Types of Virtualization - Implementation Levels of Virtualization -Virtualization Structures - Tools and Mechanisms - Virtualization of CPU, Memory, I/O Devices -Virtual Clusters and Resource management - Virtualization for Data-centre Automation.

UNIT III CLOUD INFRASTRUCTURE

Architectural Design of Compute and Storage Clouds - Layered Cloud Architecture Development- Design Challenges - Inter Cloud Resource Management - Resource Provisioning and Platform Deployment - Global Exchange of Cloud Resources

UNIT IV PROGRAMMING MODEL

Parallel and Distributed Programming Paradigms - Map Reduce, Twister and Iterative Map Reduce - Hadoop Library from Apache - Mapping Applications - Programming Support - Google App Engine, Amazon AWS - Cloud Software Environments - Eucalyptus, Open Nebula, Open Stack, Aneka, Cloud Sim. 9

SECURITY IN THE CLOUD UNIT V

Security Overview - Cloud Security Challenges and Risks - Software-as-a-Service Security -Security Governance -Risk Management – Security Monitoring – Security Architecture Design –Data Security – Application Security – Virtual Machine Security - Identity Management and Access Control - Autonomic Security. : 45

Total Contact Hours

0

9

Course Outcomes:

On completion of course students will be able to

- Compare the strengths and limitations of cloud computing •
- Identify the architecture, infrastructure and delivery models of cloud computing •
- Apply suitable virtualization concept. •
- Choose the appropriate cloud player •
- Choose the appropriate Programming Models and approach. •

Reference Books(s) / Web links

Ittl	Terence Dooks(s) / Web miks.			
1	Kai Hwang, Geoffrey C Fox, Jack G Dongarra, "Distributed and CloudComputing, From Parallel			
	Processing to the Internet of Things", Morgan Kaufmann Publishers, 2012.			
2	John W.Rittinghouse and James F.Ransome, "Cloud Computing: Implementation, Management, and Security",			
4	CRC Press, 2010.			
3	Toby Velte, Anthony Velte, Robert Elsenpeter, "Cloud Computing, A Practical Approach", TMH, 2009.			
4	Kumar Saurabh, "Cloud Computing – insights into New-Era Infrastructure", Wiley India, 2011.			
-	George Reese, "Cloud Application Architectures: Building Applications and Infrastructure in the Cloud"			
5	O'Reilly			
6	James E. Smith, Ravi Nair, "Virtual Machines: Versatile Platforms for Systems and Processes",			
U	Elsevier/Morgan Kaufmann, 2005.			
7	Katarina Stanoevska-Slabeva, Thomas Wozniak, Santi Ristol, "Grid and Cloud Computing - A Business			

	Perspective on Technology and Applications", Springer
0	Ronald L. Krutz, Russell Dean Vines, "Cloud Security – A comprehensive Guide to Secure Cloud Computing",
0	Wiley – India, 2010.
9	RajkumarBuyya, Christian Vecchiola, S.TamaraiSelvi, 'Mastering Cloud Computing", TMGH, 2013.
10	Gautam Shroff, Enterprise Cloud Computing, Cambridge University Press, 2011
11	Michael Miller, Cloud Computing, Que Publishing, 2008

со	PO1	PO2	РОЗ	PO4	PO5	PO6
C01	1	1	0	0	1	1
C02	2	2	0	0	2	2
C03	3	3	0	0	3	3
C04	3	3	0	0	3	3
C05	3	3	0	0	3	3
Average	2.4	2.4	2.4	2.4	2.4	2.4

Correlation levels 1, 2 or 3 are as defined below:

1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High)

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19203	MACHINE LEARNING TECHNIQUES	PC	3	0	0	3

Ob	Objectives:		
•	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		

		0				
UNIT I	FOUNDATIONS OF LEARNING	9				
	s of learning – learning models – geometric models – probabilistic models – logic models – groupir					
grading – le	arning versus design – types of learning – supervised – unsupervised – reinforcement – theory of lea	arning				
- feasibility	of learning - error and noise - training versus testing - theory of generalization - generalization be	ound –				
•	approximation generalization trade off – bias and variance – learning curve					
ÚNIT II	LINEAR MODELS	9				
Linear clas	sification – univariate linear regression – multivariate linear regression – regularized regression	sion –				
Logistic reg	ression – perceptrons – multilayer neural networks – learning neural networks structures – support	vector				
	soft margin SVM – going beyond linearity – generalization and over fitting – regularization – validat					
UNIT III	DISTANCE-BASED MODELS	9				
Nearest neig	hbor models – K-means – clustering around medoids – silhouttes – hierarchical clustering – k-d t	rees –				
locality sense	itive hashing – non-parametric regression – ensemble learning- bagging and random forests – boos	ting –				
meta learnir	g	•				
UNIT IV	TREE AND RULE MODELS	9				
Decision to	rees – learning decision trees – ranking and probability estimation trees – regression tr	rees –				
clustering tr	ees – learning ordered rule lists – learning unordered rule lists – descriptive rule learning – associatio	n rule				
	st-order rule learning					
UNIT V	REINFORCEMENT LEARNING	9				
	forcement learning – direct utility estimation – adaptive dynamic programming – temporal-diffe	erence				
	active reinforcement learning – exploration – learning an action utility function – Generalizat					
-	nt learning – policy search – applications in game playing – applications in robot control					
	Total Contact Hours :	45				

Coi	Course Outcomes:		
On	On completion of course students will be able to		
•	To explain theory underlying machine learning		
•	To construct algorithms to learn linear and non-linear models		
•	To implement data clustering algorithms		
•	To construct algorithms to learn tree and rule-based models		
•	To apply reinforcement learning techniques		

Ref	ference Books(s) / Web links:
1	Y. S. Abu-Mostafa, M. Magdon-Ismail, and HT. Lin, "Learning from Data", AMLBook Publishers, 2012.
2	P. Flach, "Machine Learning: The art and science of algorithms that make sense of data", Cambridge University
2	Press, 2012.
3	K. P. Murphy, "Machine Learning: A probabilistic perspective", MIT Press, 2012.
4	C. M. Bishop, "Pattern Recognition and Machine Learning", Springer, 2007
5	D. Barber, "Bayesian Reasoning and Machine Learning", Cambridge University Press, 2012.
6	M. Mohri, A. Rostamizadeh, and A. Talwalkar, "Foundations of Machine Learning", MIT Press, 2012.
7	T. M. Mitchell, "Machine Learning", McGraw Hill, 1997.

8 S. Russel and P. Norvig, "Artificial Intelligence: A Modern Approach", Third Edition, Prentice Hall, 2009.

PO CO	PO1	PO2	PO3	PO4	PO5	PO6
C01	3	2	-	2	-	1
C02	3	2	-	2	-	2
C03	2	2	1	2	-	1
C04	1	1	1	2	2	2
C05	1	1	-	2	2	1
Average	2	1.6	1	2	2	1.4

CO-PO MAPPING

Correlation levels 1, 2 or 3 are as defined below:

1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High)

Subject Code	Subject Name (Theory course)		L	Т	Р	С
DS19204	COMPUTER VISION	PC	3	0	0	3

Ob	Objectives:		
•	To review image processing techniques for computer vision		
•	To understand shape and region analysis		
•	To understand Hough Transform and its applications to detect lines, circles, ellipses		
•	To understand three-dimensional image analysis and motions.		
•	To study some applications of computer vision algorithms.		

UNIT I **IMAGE PROCESSING FOUNDATIONS**

Introduction-Image Processing Operations- Basic Image filtering operations: Noise Suppression by Gaussian Smoothing- Median Filters- Mode Filters- Rank Order Filters- The Role of Filters in Industrial Applications of Vision-Thresholding- Adaptive Thresholding-Edgedetection techniques - corner and interest point detection - mathematical morphology - Some Basic Approaches to Texture Analysis.

SHAPES AND REGIONS UNIT II

Binary shape analysis - Connectedness - Object labeling and counting - Size filtering - Distance functions -Skeletons and thinning -Other Measures for Shape Recognition - Boundary tracking procedures - Boundary Pattern Analysis- Centroidal profiles - Problems- Plot- Handling occlusion- Accuracy of boundary length measures.

UNIT III THE HOUGH TRANSFORM

Line detection- Application of Hough Transform (HT) for line detection - The Foot-of-normal method - Longitudinal line localization - Final line fitting - Using RANSAC for straight line detection Circle and Ellipse Detection: HT for circular object detection - accurate center location - speed problem - ellipse detection - Case study- Human Iris location - hole detection- Generalized Hough Transform (GHT) - Spatial matched filtering - Use of the GHT for **Ellipse Detection**

UNIT IV 3D VISION AND MOTION

3-D Vision - Methods for 3D vision - projection schemes - shape from shading - photometric stereo - Surface Smoothness- shape from texture - use of structured lighting- three dimensional object recognition schemes- Image Transformations and Camera Calibration- Motion: Optical Flow- Interpretation- Time-to-Adjacency Analysis-Difficulties- Stereo from Motion- The Kalman Filter.

UNIT V APPLICATION

Automated Visual Inspection: Process- Types- Application: Photo album - Face detection - Face recognition - Eigen faces - Active appearance and 3D shape models of faces Application- Surveillance-foreground-background separation - particle filtres - Chamfer matching- tracking- and occlusion - combining views from multiple cameras - human gait analysis Application- In-vehicle vision system: locating roadway – road markings – road signs – locating pedestrians.

Total Contact Hours

:

9

9

9

9

45

Co	Course Outcomes:		
On completion of course students will be able to			
•	Implement fundamental image processing techniques required for computer vision		
•	Perform shape analysis and able to implement boundary tracking techniques		
•	Apply Hough Transform for line, circle, and ellipse detections		
•	Apply 3D vision techniques and to implement motion related techniques		
•	Develop applications using computer vision techniques		

Ref	Reference Books(s) / Web links:			
1	E. R. Davies, "Computer & Machine Vision", Fourth Edition, Academic Press, 2012.			
2	R. Szeliski, "Computer Vision: Algorithms and Applications", Springer 2011.			
3	Simon J. D. Prince, "Computer Vision: Models, Learning, and Inference", Cambridge University Press, 2012.			
4	Mark Nixon and Alberto S. Aquado, "Feature Extraction & Image Processing for Computer Vision", Third			
4	Edition, Academic Press, 2012.			
5	D. L. Baggio et al., "Mastering OpenCV with Practical Computer Vision Projects", Packt Publishing, 2012.			

Jan Erik Solem, "Programming Computer Vision with Python: Tools and algorithms for analyzing images", O'Reilly Media, 2012.

CO-PO MAPPING

6

PO CO	PO1	PO2	PO3	PO4	PO5	PO6
C01	2	2	0	0	1	1
C02	2	2	0	0	1	1
C03	1	1	0	0	0	0
C04	1	1	0	0	1	1
C05	2	2	0	0	2	1
Average	1.6	1.6	0	0	1	0.8

Correlation levels 1, 2 or 3 are as defined below:

1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High)

Subject Code	ode Subject Name		L	Т	Р	С
AC19201	CONSTITUTION OF INDIA	MC	3	0	0	0

Ob	Objectives:	
٠	To inculcate the values enshrined in the Indian constitution.	
•	To create a sense of responsible and active citizenship.	
٠	To know about Constitutional and Non- Constitutional bodies	
•	To understand sacrifices made by the freedom fighters.	

UNIT-I INTRODUCTION

Historical Background - Constituent Assembly of India - Philosophical foundations of the Indian Constitution -Features - Basic Structure - Preamble.

UNIT-II UNION GOVERNMENT

Union and its territory - Citizenship - Fundamental Rights - Directive Principles of State Policy (DPSP) - Fundamental Duties.

Union Government: Executive, Legislature and Judiciary: President - Vice President - Prime Minister - Central Council of Ministers - Cabinet Committees - Parliament: Committees, Forums and Groups - Supreme Court.

UNIT-III STATE GOVERNMENT & UNION TERRITORIES

State Government : Executive, Legislature and Judiciary- Governor - Chief Minister - State Council of Ministers -State Legislature - High Court - Subordinate Courts -Panchayati Raj - Municipalities-Union Territories - Scheduled and Tribal Areas.

UNIT-IV **RELATIONS BETWEEN UNION AND STATES**

Relations between Union and States - Services under Union and States. Cooperative Societies - Scheduled and Tribal Areas - Finance, Property, Contracts and Suits - Trade and Commerce within Indian Territory - Tribunals. 9

UNIT-V CONSTITUTIONAL BODIES AND AMENDMENTS

Introduction to Constitutional & Non-Constitutional Bodies-Elections - Special Provisions relating to certain classes -Languages - Emergency Provisions - Miscellaneous - Amendment of the Constitution - Temporary, Transitional and Special Provisions - Short title, date of commencement, Authoritative text in Hindi and Repeals. Schedules of the Constitution of India - Appendices in the Constitution of India.

> **Total Contact Hours** : 45

9

9

Co	Course Outcomes:			
On	On completion of the course, students will be able to			
•	• Appreciate the philosophical foundations of the Indian Constitution.			
•	Understand the functions of the Indian government.			
•	Understand and abide the rules of the Indian constitution.			
•	Gain knowledge on functions of state Government and Local bodies.			
•	Gain Knowledge on constitution functions and role of constitutional bodies and amendments of constitution.			

	Text Book(s):			
1	M Lakshmikanth "Indian Polity", McGraw Hill Education, 5 th edition 2017.			
2	2 Durga Das Basu, "Introduction to the Constitution of India ", Lexis Nexis, New Delhi., 21 st edition, 2013.			
Ref	Reference Books(s) / Web links:			
1	Sharma, Brij Kishore, "Introduction to the Constitution of India", Prentice Hall of India, New Delhi, 7th edition,			
1	2015.			
2	SubhashKashyap, "Our Constitution: An Introduction to India's Constitution and Constitutional Law", National			

	Book Trust India, 1994.
3	Mahendra Prasad Singh and Himanshu Roy, "Indian Political System", Pearson India, 4 th edition, 2017.

Subject Code	Subject Name (Laboratory Course)	Category	L	Т	Р	С
DS19211	BIG DATA LABORATORY	PC	0	0	4	2

Ob	Objectives:		
•	To optimize business decisions and create competitive advantage with Big Data analytics.		
٠	To impart the architectural concepts of Hadoop and introducing map reduce paradigm.		
٠	To introduce Java concepts required for developing map reduce programs.		
•	To introduce programming tools PIG & HIVE in Hadoop echo system.		
٠	To develop Big Data applications for streaming data using Apache Spark		

List o	of Experiments			
	(i).Perform setting up and Installing Hadoop in its two operating modes:			
1.	· Pseudo distributed			
1.	· Fully distributed.			
	(ii).Use web based tools to monitor your Hadoop setup			
	Implement the following file management tasks in Hadoop:			
2.	Adding files and directories			
2.	Retrieving files			
	Deleting files			
	Run a basic Word Count Map Reduce program to understand Map Reduce Paradigm.			
3.	• Find the number of occurrence of each word appearing in the input file(s)			
	Performing a MapReduce Job for word search count (look for specific keywords in a file)			
	Stop word elimination problem:			
	Input:			
	A large textual file containing one sentence per line.			
4.	A small file containing a set of stop words (One stop word per line)			
	Output:			
	A textual file containing the same sentences of the large input file without the			
	words appearing in the small file			
	Write a Map Reduce program that mines weather data. Weather sensors collecting data every hour at many			
	locations across the globe gather large volume of log data, which is a good candidate for analysis with			
5.	MapReduce, since it is semi structured and record-oriented.			
	• Find average, max and min temperature for each year in NCDC data set?			
	• Filter the readings of a set based on value of the measurement, Output the line of input files			
6	associated with a temperature value greater than 30.0 and store it in a separate file.			
6. 7.	Install and Run Pig then write Pig Latin scripts to sort, group, join, project, and filter your data Developing applications involving concurrency			
7. 8.	Write a Pig Latin scripts for finding TF-IDF value for book dataset			
o. 9				
) 10	Install and Run Hive then use Hive to create, alter, and drop databases, tables, views, functions, and indexes.Install, Deploy & configure Apache Spark Cluster. Run apache spark applications using Scala			
10	Data analytics using Apache Spark on Amazon food dataset, find all the pairs of items frequently reviewed			
	together.			
	Write a single Spark application that:			
11.	 Transposes the original Amazon food dataset, obtaining a PairRDD of the type 			
11.	 A Counts the frequencies of all the pairs of products reviewed together; 			
	 Writes on the output folder all the pairs of products that appear more than once and their frequencies. 			
	The pairs of products must be sorted by frequency.			
	Total Contact Hours : 60			

Cou	Course Outcomes:		
•	Optimize business decisions and create competitive advantage with Big Data analytics.		
•	Imparting the architectural concepts of Hadoop and introducing map reduce paradigm.		
•	Introducing Java concepts required for developing map reduce programs.		
•	Introduce programming tools PIG & HIVE in Hadoop echo system.		
•	Developing Big Data applications for streaming data using Apache Spark.		

PO CO	PO1	PO2	PO3	PO4	PO5	PO6
C01	1	1	1	1	2	2
C02	2	2	0	0	2	2
C03	2	2	0	0	1	1
C04	3	3	1	1	2	2
C05	3	3	1	1	3	3
Average	2.2	2.2	0.6	0.6	2	2

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Laboratory Course)	Category	L	Т	Р	С
DS19212	CLOUD COMPUTING TECHNOLOGIES LABORATORY	PC	0	0	4	2

Ob	Objectives:			
•	Appreciate cloud architecture			
•	Create and run virtual machines on open source OS			
•	Implement Infrastructure, storage as a Service.			
•	Install and appreciate security features for cloud			

List	of Experiments
1	Study of Cloud Computing & Architecture
2	Virtualization in Cloud
3	Study and implementation of Infrastructure as a Service
4	Study and installation of Storage as Service
5	Implementation of identity management
6	Write a program for web feed
7	Study and implementation of Single-Sing-On
8	Securing Servers in Cloud
9	User Management in Cloud
10	Case study on Amazon EC2
11	Case study on Microsoft azure
12	Mini project
	Total Contact Hours : 60

Cou	Course Outcomes:			
•	Appreciate cloud architecture.			
•	Create and run virtual machines on open source OS.			
•	Implement Infrastructure, storage as a Service.			
•	Install and appreciate security features for cloud.			
•	Design and develop cloud computing based applications.			

РО	PO1	PO2	PO3	PO4	PO5	PO6
СО	ror	102	105	104	105	100
C01	1	1	0	0	1	1
C02	2	2	0	0	2	2
C03	3	3	0	0	3	3
C04	3	3	0	0	3	3
C05	3	3	0	0	3	3
Average	2.4	2.4	2.4	2.4	2.4	2.4

Correlation levels 1, 2 or 3 are as defined below: 1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High) No correlation : "-"

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19301	SECURITY FOR DATA SCIENCE	PC	3	0	0	3

Ob	Objectives:				
•	To understand and apply the models of Information Security.				
•	To study the Information assurance tools and methods.				
•	To study and analyze cryptographic and forensic methods.				
•	To analyze and simulate the network and application security.				
•	To explore the nature and logic behind the cyber security threats as an ethical hacker.				

UNIT I	AUTHENTICATION APPLICATIONS	9
Authenticati	on Applications : Kerberos - X.509 Authentication service - Public-key Infrastructure Electronic	: Mail
Security : Pr	retty Good Privacy – S/MIME	
UNIT II	NETWORK SECURITY	8
IP Security	: IP Security Overview- IP security Architecture - Authentication Header - Encapsulating Se	curity
Payload – 0	Combining Security Associations - Key Management Web Security : Web security Considerat	ions -
Secure Sock	tet Layer and Transport Layer Security- Secure Electronic Transactions.	
UNIT III	IP AND WEB SECURITY	9
Network see	curity - Intrusion Prevention, detection and Management - Firewall - Ecommerce Security - Con	nputer

Forensics - Security for VPN and Next Generation Networks DATA PRIVACY AND ANONYMIZATION UNIT IV 10 Understanding Privacy: Social Aspects of Privacy Legal Aspects of Privacy and Privacy Regulations Effect of Database and Data Mining technologies on privacy challenges raised by new emerging technologies such RFID, biometrics, etc., Privacy Models

Introduction to Anonymization, Anonymization models: K-anonymity, 1-diversity, t-closeness, differential privacy Database as a service

DATA PRIVACY FOR DATA SCIENCE UNIT V

Using technology for preserving privacy: Statistical Database security Inference Control Secure Multi-party

computation and Cryptography Privacy-preserving Data mining Hippocratic databases Emerging Applications: Social Network Privacy, Location Privacy, Query Log Privacy, Biomedical Privacy

Total Contact Hours

9

9

: 45

Course Outcomes: On completion of course students will be able to

- Identify the information security models and their characteristics. •
- Analyse the different types of cryptographic and forensic methods. ٠
- Study the network security issues. •
- Understand the privacy and anonymoization ٠
- Identify and solve different cyber security threats

Ref	Cerence Books(s) / Web links:					
1	William Stallings, "Cryptography and Network Security: Principles and Practice", 6 th Edition, PHI, 2014.					
2	Michael E. Whitman and Herbert J Mattord, "Principles of Information Security", 6 thedition, Vikas Publishing					
4	House, 2017.					
2	Bill Nelson, Amelia Phillips, F.Enfinger and Christopher Stuart, "Guide to Computer Forensics and					
3	Investigations, 4 th ed., Thomson Course Technology, 2010					
4	Matt Bishop, "Computer Security: Art and Science", 1 st edition, Addison-Wesley Professional, 2015.					
5	Privacy-Preserving Data Mining- Models and Algorithms, Charu C Aggarwal, Yu Philips, S., Springer					
6	Principles of Information Security, Information SecurityProfessional - Michael E. Whitman and Herbert J.					

PO CO	PO1	PO2	PO3	PO4	PO5	PO6
C01	-	-	3	2	2	2
C02	-	1	2	3	2	2
C03	-	-	2	2	-	2
C04	-	-	-	-	2	2
C05	-	-	-	2	2	2
Average	-	1	1.4	1.8	1.6	2

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19P21	ADVANCED DATABASE TECHNOLOGY	PE	3	0	0	3

Obj	jectives:
•	To understand the design of databases
•	To acquire knowledge on parallel databases and its applications.
•	To acquire knowledge on distributed databases and its applications.
•	To understand the usage and applications of Object Oriented and Intelligent databases and emerging databases
	like Mobile, XML, Cloud and Big Data
•	To learn the special databases like Multimedia, text/document databases

UNIT I PARALLEL AND DISTRIBUTED DATABASES

 Database System Architectures: Centralized and Client-Server Architectures – Server System Architectures – Parallel

 Systems- Distributed Systems – Parallel Databases: I/O Parallelism – Inter and Intra Query Parallelism – Inter and

 Intra operation Parallelism – Design of Parallel Systems Distributed Database Concepts - Distributed Data Storage –

 Distributed Transactions – Commit Protocols – Concurrency Control – Distributed Query Processing – Case Studies

 UNIT II
 INTELLIGENT DATABASES

 Active Databases: Syntax and Semantics (Starburst, Oracle, DB2)- Taxonomy- Applications- Design Principles for

 Active Rules- Temporal Databases: Overview of Temporal Databases TSQL2- Deductive Databases-Recursive

 Queries in SQL- Spatial Databases- Spatial Data Types - Spatial Relationships- Spatial Data Structures-Spatial Access

Methods- Spatial DB Implementation.

UNIT III XML DATABASES

XML Databases: XML Data Model – DTD – XML Schema – XML Querying – Web Databases – Open Database Connectivity.

UNIT IV MOBILE DATABASES

 Mobile Databases: Location and Handoff Management - Effect of Mobility on Data Management - Location

 Dependent Data Distribution - Mobile Transaction Models Concurrency Control - Transaction Commit Protocols

 UNIT V
 MULTIMEDIA DATABASES

 9

Multidimensional Data Structures – Image Databases – Text / Document Databases – Video Databases – Audio Databases – Multimedia Database Design.

Total Contact Hours:60

9

9

Course Outcomes:

On completion of course students will be able to

~	completion of course students will be usie to
•	Understand the design of databases
•	Acquire knowledge on parallel databases and its applications.
•	Acquire knowledge on distributed databases and its applications.
•	Explain the usage and applications of object oriented and intelligent databases and Understand the emerging databases like mobile, xml, cloud and big data
•	Explain the special databases like Multimedia, text/document databases

Reference Books(s) / Web links:

INCI	terence books(s) / web miks.			
1	Carlo Zaniolo, Stefano Ceri, Christos Faloutsos, Richard T.Snodgrass, V.S.Subrahmanian, Roberto Zicari, —Advanced Database Systems ^{II} , Morgan Kaufmann publishers,2006.			
1	-Advanced Database Systems, Morgan Kaufmann publishers, 2006.			
2	C.J.Date, A.Kannan, S.Swamynathan, -An Introduction to Database Systems, Eighth Edition, Pearson			
2	Education, 2012.			
3	Bipin Desai,". An Introduction to Database System. Galgotia Publications Pvt Ltd-New Delhi 2012.			
4	Henry F Korth, Abraham Silberschatz, S. Sudharshan, -Database System Concepts, Sixth Edition, McGraw			
4	Hill, 2011.			
5	R. Elmasri, S.B. Navathe, -Fundamentals of Database Systems, Sixth Edition, Pearson Education/Addison			
3	Wesley, 2010.			

РО	PO1	PO2	PO3	PO4	PO5	PO6
СО	101	102	100	101	100	100
C01	3	2	0	0	1	1
C02	3	3	0	0	1	1
C03	3	3	0	0	1	1
C04	2	2	0	0	2	2
C05	2	2	0	0	2	2
Average	2.6	2.4	0	0	1.4	1.4

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19P22	VIDEO ANALYTICS	PE	3	0	0	3

Ob	Objectives:				
•	To know the fundamental concepts of big data and analytics				
•	To learn various techniques for mining data streams				
•	To acquire the knowledge of extracting information from surveillance videos.				
•	• To learn Event Modelling for different applications.				
•	To understand the models used for recognition of objects in videos.				

UNIT I **INTRODUCTION TO BIG DATA & DATA ANALYSIS**

Introduction to Big Data Platform - Challenges of Conventional systems - Web data- Evolution of Analytic scalability- analytic processes and tools- Analysis Vs Reporting- Modern data analytic tools- Data Analysis: Regression Modeling- Bayesian Modeling- Rule induction. 9

UNIT II MINING DATA STREAMS

Introduction to Stream concepts- Stream data model and architecture - Stream Computing- Sampling data in a Stream-Filtering Streams- Counting distinct elements in a Stream- Estimating moments- Counting oneness in a window-Decaying window- Real time Analytics platform(RTAP) applications- case studies.

UNIT III VIDEO ANALYTICS 9 Introduction- Video Basics - Fundamentals for Video Surveillance- Scene Artifacts- Object Detection and Tracking: Adaptive Background Modelling and Subtraction- Pedestrian Detection and TrackingVehicle Detection and Tracking-Articulated Human Motion Tracking in Low-Dimensional Latent Spaces

BEHAVIOURAL ANALYSIS & ACTIVITY RECOGNITION UNIT IV

Behavioural Analysis- Human Activity Recognition-Complex Activity Recognition- Activity modelling using 3D shape, Video summarization, shape based activity models- Suspicious Activity Detection 9

UNIT V **HUMAN FACE RECOGNITION & GAIT ANALYSIS**

Overview of Recognition algorithms - Human Recognition using Face: Face Recognition from still images, Face Recognition from video, Evaluation of Face Recognition Technologies- Human Recognition using gait: HMM Framework for Gait Recognition, View Invariant Gait Recognition, Role of Shape and Dynamics in Gait Recognition **Total Contact Hours** : 45

Co	Course Outcomes:		
On completion of course students will be able to			
•	Work with big data platform and its analysis techniques.		
•	Design efficient algorithms for mining the data from large volumes.		
٠	Work with surveillance videos for analytics.		
•	Design of optimization algorithms for better analysis and recognition of objects in a scene.		

Model a framework for Human Activity Recognition.

Ref	Reference Books(s) / Web links:					
1	AnandRajaraman and Jeffrey David Ullman, Mining of Massive Datasets, Cambridge University Press, 2012.					
2	Michael Berthold, David J.Hand, Intelligent Data Analysis, Springer, 2007.					
2	Rama Chellappa, Amit K.Roy-Chowdhury, Kevin Zhou.S, "Recognition of Humans and their Activities using					
3	Video", Morgan&Claypool Publishers, 2005.					
4	Yunqian Ma, Gang Qian, "Intelligent Video Surveillance: Systems and Technology", CRC Press (Taylor and					
4	Francis Group), 2009.					

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РО	PO1	PO2	PO3	PO4	PO5	PO6
СО	101	102	100	101	100	100
C01	3	2	-	2	1	1
C02	2	3	-	-	2	2
C03	3	3	1	-	2	2
C04	1	3	-	1	2	2
C05	2	2	1	_	3	3
Average	2.2	2.6	0.4	0.6	2	2

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Theory course)	Category	L	Т	P	С
DS19P23	DEEP LEARNING	PE	3	0	0	3

Ob	jectives:
•	To understand the fundamental principles, theory and approaches for learning with deep neural networks
•	To learn the main variants of deep learning (such convolutional and recurrent architectures), and their typical applications
•	To understand the key concepts, issues and practices when training and modelling with deep architectures
•	To understand how deep learning fits within the context of other ML approaches
٠	To learn deep learning algorithms in application of choice using tensor flow.

UNIT I	INTRODUCTION TO DEEP LEARNING	9		
Linear Regi	ression -Nonlinear Regression- Logistic Regression Activation			
UNIT II CONVOLUTIONAL NEURAL NETWORKS (CNN)				
CNN Histor	ry- Understanding CNNs- CNN Application			
UNIT III	RECURRENT NEURAL NETWORKS (RNN)	9		
	N Model Long Short-Term memory (LSTM) Recursive Neural Tensor Network Theory Recurrent	Neural		
	N Model Long Short-Term memory (LSTM) Recursive Neural Tensor Network Theory Recurrent	Neural		
Intro to RN	N Model Long Short-Term memory (LSTM) Recursive Neural Tensor Network Theory Recurrent	Neural		
Intro to RN Network M UNIT IV	N Model Long Short-Term memory (LSTM) Recursive Neural Tensor Network Theory Recurrent odel			
Intro to RN Network M UNIT IV	N Model Long Short-Term memory (LSTM) Recursive Neural Tensor Network Theory Recurrent odel UNSUPERVISED LEARNING			
Intro to RN Network M UNIT IV Application UNIT V	N Model Long Short-Term memory (LSTM) Recursive Neural Tensor Network Theory Recurrent odel UNSUPERVISED LEARNING s of Unsupervised Learning-Restricted Boltzmann Machine-Collaborative Filtering with RBM	9		

Course Outcomes:

On completion of course students will be able to

• Understand the fundamental principles, theory and approaches for learning with deep neural networks

- Implement the main variants of deep learning (such convolutional and recurrent architectures), and their typical applications
- Understand the key concepts, issues and practices when training and modeling with deep architectures.
- Understand how deep learning fits within the context of other ML approaches.
- Apply deep learning algorithms in application of choice using tensor flow.

Ref	Reference Books(s) / Web links:		
1	Ian Goodfellow, YoshuaBengio and Aaron Courville, "Deep Learning", The MIT Press		
2	Rajiv Chopra, Deep Learning: A Practical Approach, Khanna Publication		
3	MOOC, Deep Learning By Google, <u>https://in.udacity.com/course/deep-learningud730</u>		
4	MOOC, Deep Learning https://www.coursera.org/specializations/deep-learning		

РО	PO1	PO2	PO3	PO4	PO5	PO6
СО	101	102	105	104	105	100
C01	2	3	1	2	3	-
C02	3	3	-	2	3	-
C03	3	3	-	3	3	-
C04	3	3	1	3	3	1
C05	3	3	1	3	3	1
Average	2.8	3	0.6	2.6	3	0.4

Correlation levels 1, 2 or 3 are as defined below:

1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High)

No correlation : "-"

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19P24	FINITE ELEMENTS: THEORY AND ALGORITHMS	PE	3	0	0	3

Ob	Objectives:	
•	To understand the Sobolev spaces and finite element spaces	
•	To learn the Biharmonic equation	
•	To study and understand the Parabolic problems	
•	To analyse the Systems in solid and fluid mechanics	
•	To learn the implementation of the finite element method	

UNIT I	SOBOLEV SPACES AND FINITE ELEMENT SPACES	9	
Banach and Hilbert spaces - Weak derivatives - Sobolev spaces - Elliptic scalar problems - Simplices and barycentric			
coordinates	- Simplicial finite elements and local spaces - Construction of finite element spaces - The conce	ept of	
	te elements: Affine mappings - Finite elements on rectangular and brick meshes - Mapped finite eler	nents:	
General bije	ctive mappings - Mapped Q k finite element - Isoparametric finite elements.		
UNIT II	BIHARMONIC EQUATION	9	
	tion formulas - Affine equivalent finite elements - Canonical interpolation - Local and global interpo oved L 2 error estimates by duality - Interpolation of less smooth functions - Deflection of a thin cla		
plate - Wea	k formulation of the biharmonic equation - Conforming finite element methods - Nonconforming	finite	
element met	hods.		
		9	
UNIT III	PARABOLIC PROBLEMS	-	
*	n of energy - A general parabolic problem of second order - Weak formulation of initial boundary	-	
Conservatio		value	
Conservatio	n of energy - A general parabolic problem of second order - Weak formulation of initial boundary	value	
Conservatio problems - S	n of energy - A general parabolic problem of second order - Weak formulation of initial boundary	value	
Conservatio problems - S problems. UNIT IV	n of energy - A general parabolic problem of second order - Weak formulation of initial boundary Semidiscretization by finite elements - Time discretization - Finite elements for high-dimensional para	value abolic 9	
Conservatio problems - S problems. UNIT IV Linear elast problem - C	n of energy - A general parabolic problem of second order - Weak formulation of initial boundary Semidiscretization by finite elements - Time discretization - Finite elements for high-dimensional para SYSTEMS IN SOLID AND FLUID MECHANICS icity - Mindlin–Reissner plate - Conservation of mass and momentum - Weak formulation of the Store problem - Nonconforming discretizations of the Stokes problem - Nonconforming discretizat	value abolic 9 Stokes olem -	
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Conservatio problems - S problems. UNIT IV Linear elast problem - C The noncon	n of energy - A general parabolic problem of second order - Weak formulation of initial boundary Semidiscretization by finite elements - Time discretization - Finite elements for high-dimensional para SYSTEMS IN SOLID AND FLUID MECHANICS icity - Mindlin–Reissner plate - Conservation of mass and momentum - Weak formulation of the S conforming discretizations of the Stokes problem - Nonconforming discretizations of the Stokes prob forming Crouzeix–Raviart element - Further inf–sup stable finite element pairs - Equal order stab nts - Navier–Stokes problem with mixed boundary conditions - Time discretization and linearization	value abolic 9 Stokes olem - pilized	
Conservatio problems - S problems. UNIT IV Linear elast problem - C The noncon finite element	n of energy - A general parabolic problem of second order - Weak formulation of initial boundary Semidiscretization by finite elements - Time discretization - Finite elements for high-dimensional para SYSTEMS IN SOLID AND FLUID MECHANICS icity - Mindlin–Reissner plate - Conservation of mass and momentum - Weak formulation of the S conforming discretizations of the Stokes problem - Nonconforming discretizations of the Stokes prob forming Crouzeix–Raviart element - Further inf–sup stable finite element pairs - Equal order stab nts - Navier–Stokes problem with mixed boundary conditions - Time discretization and linearization	value abolic 9 Stokes olem - pilized	
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Conservatio problems - S problems. UNIT IV Linear elast problem - C The noncon finite elemen Navier–Stok UNIT V Mesh handli	n of energy - A general parabolic problem of second order - Weak formulation of initial boundary Semidiscretization by finite elements - Time discretization - Finite elements for high-dimensional para SYSTEMS IN SOLID AND FLUID MECHANICS icity - Mindlin–Reissner plate - Conservation of mass and momentum - Weak formulation of the Stonforming discretizations of the Stokes problem - Nonconforming discretizations of the Stokes problem forming Crouzeix–Raviart element - Further inf–sup stable finite element pairs - Equal order stab nts - Navier–Stokes problem with mixed boundary conditions - Time discretization and linearization ces problem. IMPLEMENTATION OF THE FINITE ELEMENT METHOD ing and data structure - Numerical integration - Sparse matrix storage - Assembling of system matrices s - Inclusion of boundary conditions - Solution of the algebraic systems - Object-oriented	value abolic 9 Stokes blem - bilized of the 9 es and	

Co	Course Outcomes:		
On	On completion of course students will be able to		
•	Understand the Sobolev spaces and finite element spaces.		
•	Apply the Biharmonic equation.		
•	Solve the Parabolic problems.		
•	Analyse the Systems in solid and fluid mechanics.		
•	Implement the finite element method.		

Reference Books(s) / Web links:

	SasikumaarGanesan, Lutz Tobiska, "Finite Elements", Cambridge University Press
2	T R Chandrupatla and A D Belegundu," Introduction to Finite Elements in Engineering"

РО	PO1	PO2	PO3	PO4	PO5	PO6
СО	101	102	105	104	105	100
C01	2	1	3	1	2	1
C02	2	1	3	1	2	1
C03	2	1	3	1	2	1
C04	2	1	3	1	2	1
C05	2	1	3	1	2	1
Average	2	1	3	1	2	1

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19P25	INFORMATION RETRIEVAL TECHNIQUES	PE	3	0	0	3

Ob	jectives:
•	To understand the basics of information retrieval with pertinence to modeling, query operations and indexing
•	To get an understanding of machine learning techniques for text classification and clustering.
•	To understand the various applications of information retrieval giving emphasis to multimedia IR, web search
•	To understand the concepts of digital libraries.
•	To know about Indexing and Searching.

UNIT I **INTRODUCTION MOTIVATION**

Basic Concepts - Practical Issues - Retrieval Process - Architecture - Boolean Retrieval - Retrieval Evaluation - Open Source IR Systems-History of Web Search - Web Characteristics-The impact of the web on IR --- IR Versus Web Search-Components of a Search engine

UNIT II MODELING

Taxonomy and Characterization of IR Models - Boolean Model - Vector Model - Term Weighting - Scoring and Ranking -Language Models - Set Theoretic Models - Probabilistic Models - Algebraic Models - Structured Text Retrieval Models - Models for Browsing

UNIT III INDEXING

Static and Dynamic Inverted Indices - Index Construction and Index Compression. Searching - Sequential Searching and Pattern Matching. Query Operations -Query Languages - Query Processing - Relevance Feedback and Query Expansion - Automatic Local and Global Analysis - Measuring Effectiveness and Efficiency 9

UNIT IV CLASSIFICATIONAND CLUSTERING

Text Classification and Naïve Bayes - Vector Space Classification - Support vector machines and Machine learning on documents. Flat Clustering - Hierarchical Clustering -Matrix decompositions and latent semantic indexing - Fusion and Meta learning

UNIT V SEARCHINGTHE WEB

Searching the Web -Structure of the Web -IR and web search - Static and Dynamic Ranking - Web Crawling and Indexing - Link Analysis - XML Retrieval Multimedia IR: Models and Languages - Indexing and Searching Parallel and Distributed IR - Digital Libraries

- **Total Contact Hours**
 - :

9

9

9

9

45

	Course Outcomes:						
On	On completion of course students will be able to						
•	Build an Information Retrieval system using the available tools						
•	Identify and design the various components of an Information Retrieval system						
•	Apply machine learning techniques to text classification and clustering which is used for efficient Information						
•	Retrieval						
٠	Design an efficient search engine and analyze the Web content structure.						
٠	Gained knowledge about Support Vector Machine.						

Ref	ference Books(s) / Web links:
1	Christopher D. Manning, PrabhakarRaghavan, HinrichSchutze, -Introduction to Information Retrievall,
1	Cambridge University Press, First South Asian Edition, 2008.
	Ricardo Baeza - Yates, Berthier Ribeiro - Neto, -Modern Information Retrieval: The concepts and
2	Technology behind Searchl (ACM Press Books), Second Edition, 2011.
2	Stefan Buttcher, Charles L. A. Clarke, GordonV. Cormack, -Information Retrieval Implementing and
3	Evaluating Search Engines, The MIT Press, Cambridge, Massachusetts London, England, 2010.

РО	PO1	PO2	PO3	PO4	PO5	PO6
CO						
C01	3	2	1	-	2	-
C02	3	2	-	-	2	-
C03	3	2	1	1	2	1
C04	3	2	-	1	2	1
C05	3	2	1	1	3	1
Average	3	2	1	1	2.2	1

Correlation levels 1, 2 or 3 are as defined below:

Subjec	et Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS1	9P26	NATURAL LANGUAGE PROCESSING	PE	3	0	0	3

Ob	Objectives:					
•	To provide the understanding of Natural Language Processing and challenges involved in that area.					
•	To provide the student with knowledge of various levels of analysis involved in NLP					
•	To understand language modeling,					
•	To gain knowledge in automated natural language generation and machine translation					

OVERVIEW AND LANGUAGE MODELING UNIT I

Overview: Origins and challenges of NLP-Language and Grammar-Processing Indian Languages-NLP Applications-Information Retrieval.LanguageModeling: Introduction-Various Grammar-based Language Models-Statistical Language Model

UNIT II WORD LEVEL AND SYNTACTIC ANALYSIS

Word Level Analysis: Introduction- Regular Expressions-Finite-State Automata-Morphological Parsing-Spelling Error Detection and correction-Words and Word classes-Part-of Speech Tagging. Syntactic Analysis: Introduction-Contextfree Grammar-ConstituencyParsing-Probabilistic Parsing

SEMANTIC ANALYSIS AND DISCOURSE PROCESSING UNIT III

Semantic Analysis: Introduction-Meaning Representation-Lexical SemanticsAmbiguity-Word Sense Disambiguation.Discourse Processing: Introduction- cohesion-Reference ResolutionDiscourse Coherence and Structure 0

UNIT IV NATURAL LANGUAGE GENERATION AND MACHINE TRANSLATION

Natural Language Generation: Introduction-Architecture of NLG SystemsGeneration Tasks and Representations-Application of NLG.Machine Translation: Introduction-Problems in Machine TranslationCharacteristics of Indian Languages- Machine Translation Approaches-Translation involving Indian Languages

INFORMATION RETRIEVAL AND LEXICAL RESOURCES UNIT V

Information Retrieval: Introduction-Design features of Information Retrieval Systems-Classical, Non-classical, Alternative Models of Information Retrieval - EvaluationLexical Resources: Introduction-WordNet-FrameNet-Stemmers-POS TaggerResearch Corpora

Total Contact Hours 45 :

8

9

10

9

Co	Course Outcomes:						
On	On completion of course students will be able to						
•	Analyse the challeges and model the natural language						
•	Perform word level and syntactic analysis						
•	Do semantic analysis and discourse processing						
•	Automate natural language generation and machine translation						
٠	Evaluate information retrieval and laexical resources						
Ref	ference Books(s) / Web links:						

ICI	terence books(s) / web miks.
1	Tanveer Siddiqui, U.S. Tiwary, "Natural Language Processing and Information Retrieval", Oxford University
1	Press, 2008.
2	Daniel Jurafsky and James H Martin, "Speech and Language Processing: An introduction to Natural Language
4	Processing, Computational Linguistics and Speech Recognition", Prentice Hall, 2nd Edition, 2008.
3	James Allen, Bejamin/cummings, "Natural Language Understanding", 2nd edition, 1995. C

РО	PO1	PO2	PO3	PO4	PO5	PO6
СО					100	200
C01	3	2	1	1	1	1
C02	3	2	1	1	2	1
C03	3	2	1	1	1	1
C04	3	2	1	1	3	1
C05	3	2	1	1	1	1
Average	3	2	1	1	1.6	1

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19P27	SOCIAL NETWORK ANALYSIS	PE	3	0	0	3

Ob	Objectives:	
•	Describe the concepts and measures in social network analysis	
•	Explain the modelling and visualization of social networks	
•	Learn the Text preprocessing and Text summarization techniques	
•	Study the different types of Text classification algorithms	
	Learn the various Text Clustering algorithms	

UNIT I	INTRODUCTION TO SOCIAL NETWORK ANALYSIS	9
Introduction	to Web - Limitations of current Web – Development of Semantic Web – Emergence of the Social	Web -
Network an	alysis - Development of Social Network Analysis - Key concepts and measures in network ana	lysis -
Electronic s	ources for network analysis - Electronic discussion networks, Blogs and online communities, Web	based
networks.		
UNIT II	MODELLING AND VISUALIZING SOCIAL NETWORKS	9
	Online Social Networks, Graph Representation - Centrality- Clustering - Node-Edge Dia	
	Social Networks with Matrix-Based Representations Matrix -Link Diagrams, Hybrid Representation	
	and aggregating social network data - Ontological representation of social individuals, Ontol	logical
	on of social relationships.	
UNIT III	TEXT MINING IN SOCIAL MEDIA	9
	-Text Preprocessing- Tokenization, Stop word removal, Stemming; Information Extraction from	
	tity Recognition, Relation Extraction, Unsupervised Information Extraction; Text Summari	
	Topic Representation Approaches, Influence of Context; Dimensionality Reduction and Topic Mode	-
UNIT IV	TEXT CLASSIFICATION ALGORITHMS	9
	-Feature Selection for Text Classification-Gini Index, Information Gain, Mutual Information,χ2-S	
	ansformation Methods: Supervised LSI, Supervised Clustering for Dimensionality Reduction,	
	t Analysis, Generalized Singular Value Decomposition, Interaction of Feature Selection	
	on; Probabilistic and Naive Bayes Classifiers- Bernoulli Multivariate Model, Multinomial Distribution for Text Classification of Linked and Web Determined and Web De	
	delling for Text Classification; Proximity-based Classifiers; Classification of Linked and Web Data; for Text Classification- Classifier Ensemble Learning, Data Centered Methods: Boosting and Ba	
-	Specific Measures of Accuracy	igging,
UNIT V	TEXT CLUSTERING ALGORITHMS	9
<u> </u>	-Feature Selection and Transformation Methods for Text Clustering- Feature Selection Methods, LS	-
	ods, Non-negative Matrix Factorization; Distance-based Clustering Algorithms- Agglomerative and	1-
		1
	Clustering Algorithms, Distance-based Partitioning Algorithms, A Hybrid Approach: The Scatter-G	
	ord and Phrase-based Clustering- Clustering with Frequent Word Patterns, Leveraging Word Clusters	
Document (Clusters, Co-clustering Words and Documents, Clustering with Frequent Phrases; Online Clustering v	with
Text Stream	s; Clustering Text in Networks; Semi-Supervised Clustering.	
	Total Contact Hours :	45

Cou	Course Outcomes:	
On	On completion of course students will be able to	
•	Interpret the measures in social network analysis	
•	Illustrate and design social network	
•	Apply the various preprocessing techniques on text	
•	Implement various Text Classification techniques	
•	Compare and contrast Text classification and Text Clustering	

TEXT BOOKS

- 1. Peter Mika, "Social networks and the Semantic Web", Springer, 1 st edition 2007.
- 2. Borko Furht, "Handbook of Social Network Technologies and Applications", Springer, 1st edition, 2010.
- 3. Charu. Aggarwal, ChengxiangZhai, "Mining Text data", Springer, 2012.

PO CO	PO1	PO2	PO3	PO4	PO5	PO6
C01	3	3	-	2	3	2
C02	3	3	-	2	3	2
C03	3	3	-	2	3	2
C04	3	3	-	2	3	2
C05	3	3	-	2	3	2
Average	3	3	-	2	3	2

Correlation levels 1, 2 or 3 are as defined below:

1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High)

No correlation : "-"

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19P28	RECOMMENDATION SYSTEMS	PE	3	0	0	3

U D	ojectives:
•	To study the basic concepts and functions of recommendation systems.
•	To learn about various model based and neighborhood collaborative filters.
•	To study various components of content and knowledge – based systems.
•	To learn ensemble and hybrid – based systems.
•	Understand and evaluate the recommendation systems based on various metrics.

UNIT-I	INTRODUCTION TO RECOMMENDER SYSTEMS
Introduction	- Goals - Basic Models - Collaborative Filtering Model - Content Based Recommender Sy
Knowledge	Based Recommender Systems – Demographic Recommender Systems – Hybrid and Ensembl

ystems -- Demographic Recommender Systems - Hybrid and Ensemble Based Based Recommender Systems Recommender Systems - domain specific challenges - Context Based Recommender Systems - Time-Sensitive Recommender Systems - Location Based Recommender Systems - Social Recommender Systems - Advanced Topics and Applications.

UNIT-II	NEIGHBORHOOD AND MODEL-BASED COLLABORATIVE FILTERING	9
Neighborhoo	od: Introduction - Key Properties - Predicting Ratings - Clustering - Dimensionality Reduct	ion –
Regression	Modeling - Graph Models. Model: Introduction - Decision and Regression Trees - Rule	Based
Collaborativ	e Filtering - Naive Baye's Collaborative Filtering - Latent Factor Models - Integrating Factorizatio	on and
Neighborhoo	od Models.	

UNIT-III	CONTENT AND KNOWLEDGE-BASED RECOMMENDER SYSTEMS	9
Content: Intr	roduction - Basic Components - Preprocessing and Feature Extraction - Learning User Profile	es a
Filtering – C	Content Based Versus Collaborative Recommendations - Using Content Based Models for Collaborative	orati
Filtering. Kr	nowledge: Introduction - Constraint Based Recommender Systems - Case Based Recommender	ders

and tive ·s – Persistent Personalization in Knowledge-Based Systems.

UNIT-IV ENSEMBLE-BASED AND HYBRID RECOMMENDER SYSTEMS

Introduction - Ensemble Methods from the Classification Perspective - Weighted Hybrids - Switching Hybrids -Cascade Hybrids - Feature Augmentation Hybrids - Meta Level Hybrids - Feature Combination Hybrids - Mixed Hybrids.

EVALUATING RECOMMENDER SYSTEMS **UNIT-V**

Evaluating: Introduction - Evaluation Paradigms - General Goals of Evaluation Design: Accuracy, Coverage, Confidence and Trust, Novelty, Serendipity, Diversity, Robustness and Stability and Scalability- Design Issues in Offline Recommender Evaluation - Accuracy Metrics in Offline Evaluation - Limitations of Evaluation Measures -Case study on Recommender Systems.

> **Total Contact Hours** 45 :

9

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Co	Course Outcomes:			
On	completion of the course, students will be able to			
Describe the basic concepts and various techniques for making recommendations.				

•	Design the key properties of neighborhood and model-based collaborative filtering.
•	Design recommendation system based on content and knowledge.
•	Develop a new hybrid recommender systems.
•	Evaluate recommender systems on the basis of measuring metrics such as coverage, confidence and trust.

Tex	Text Book:					
	Charu C. Aggarwal, "Recommender Systems", Springer Cham Heidelberg New York Dordrecht London					
1	©Springer International Publishing Switzerland 2016.					
Ref	ference Books(s) / Web links:					
1	Ricci F., Rokach L., Shapira D., Kantor B.P., Recommender Systems Handbook, Springer, 2011.					
2	Manouselis N., Drachsler H., Verbert K., Duval E., Recommender Systems for Learning, Springer, 2013.					
	Jannach D., Zanker M. and FelFering A., Recommender Systems: An Introduction, Cambridge University Press					
3	(2011), 1st ed.					

РО	PO1	PO2	PO3	PO4	PO5	PO6
СО	101	F02	105	104	105	100
C01	3	3	-	2	3	2
C02	3	2	-	1	2	2
C03	3	2	-	1	2	2
C04	3	2	-	1	2	2
C05	2	3	-	-	2	1
Average	2.8	2.4	-	1	2.2	1.8

Correlation levels 1, 2 or 3 are as defined below: 1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High) No correlation : "-"

Subject Code	Subject Name (Theory course)		L	Т	Р	С
DS19P31	HEALTH CARE ANALYTICS	PE	3	0	0	3

Ob	Objectives:			
٠	To explore the various forms of electronic health care information.			
٠	To learn the techniques adopted to analyse health care data.			
	To understand the predictive models for clinical data			

To understand the predictive models for clinical data

INTRODUCTION UNIT I

Introduction to Healthcare Data Analytics- Electronic Health Records- Components of EHR- Coding Systems-Benefits of EHR- Barrier to Adopting EHRChallenges- Phenotyping Algorithms.

UNIT II ANALYSIS

Biomedical Image Analysis- Mining of Sensor Data in Healthcare- Biomedical Signal Analysis- Genomic Data Analysis for Personalized Medicine.

UNIT III ANALYTICS

Natural Language Processing and Data Mining for Clinical Text- Mining the Biomedical- Social Media Analytics for Healthcare.

ADVANCED DATA ANALYTICS UNIT IV

Advanced Data Analytics for Healthcare- Review of Clinical Prediction Models- Temporal Data Mining for Healthcare Data- Visual Analytics for Healthcare- Predictive Models for Integrating Clinical and Genomic Data-Information Retrieval for Healthcare- Privacy-Preserving Data Publishing Methods in Healthcare. 9

UNIT V APPLICATIONS

Applications and Practical Systems for Healthcare- Data Analytics for Pervasive Health- Fraud Detection in Healthcare- Data Analytics for Pharmaceutical Discoveries- Clinical Decision Support Systems- Computer-Assisted Medical Image Analysis Systems- Mobile Imaging and Analytics for Biomedical Data.

Total Contact Hours : 45

9

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Course Outcomes: On completion of course students will be able to Analyze health care data using appropriate analytical techniques. • • Apply analytics for decision making in healthcare services. Apply data mining to integrate health data from multiple sources and develop efficient clinical decision support systems.

Text Books(s) / Web links:

1	Chandan K. Reddy and Charu C Aggarwal, "Healthcare data analytics", Taylor & Francis, 2015
2	Hui Yang and Eva K. Lee, "Healthcare Analytics: From Data to Knowledge to Healthcare Improvement, Wiley, 2016.

Text Books(s) / Web links:

	At DOORS(5) / Web minds
1	Trevor L. Strome (2013). Healthcare Analytics for Quality and Performance Improvement. John Wiley & Sons,
1	Inc
2	Principles of Health Interoperability HL7 and SNOMED (Health Information Technology Standards), Springer
4	Publication by Tim Benson
3	Health Service Marketing – A Practitioners Guide - Richard K. T Springer New York.
4	Wellness Management A Lifestyle Approach for Health, Fitness and Energy - Rajasekhar Kali Venkata - Notion
4	press
5	Vittinghoff E., Glidden, D.V., Shiboski, S.C. and McCulloch, C.E. (2005) Regression Methods in Biostatistics.
6	DeVellis, Robert F. (2012). Scale Development: Theory and Applications; Third Edition. Thousand Oaks, CA:
6	Sage Publications

РО						
со	PO1	PO2	PO3	PO4	PO5	PO6
C01	2	3	-	1	3	2
C02	2	3	-	1	3	2
C03	2	3	-	1	3	2
C04	2	3	-	1	3	2
C05	2	3	-	1	3	2
Average	2	3	-	1	3	2

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19P32	FINANCIAL AND RISK ANALYTICS	PE	3	0	0	3

Ob	Objectives:		
•	To understand the basics of Financial and Risk Analytics		
•	To learn the basics of Financial Research and Operational Risk		
٠	To study the financial decision making and Risk Management		
•	To learn the Functional Framework for Evaluating Financial Visualization Products.		
•	To learn and implement preserving digital records		

UNIT-I INTRODUCTION TO FINANCIAL AND RISK ANALYTICS 9

Records and Information Management for Financial Analysis and Risk Management: Financial Decision Process: Theory and Practice, Terminology, Governance, Analytics, Long-Term Digital Preservation. Monitoring Financial Stability in a Complex World :Introduction ,Legacy Financial Supervision ,Firm-Level Supervision and Disintermediation ,Financial Innovation and the Complexity of Data Management, Scalability of Data Management, Systemic Supervision and the Network of Counterparty Claims , Networks and Information ,An Example: Rehypothecation of Repo Collateral, Implications for Supervisory Implementation.

On Operational Risk, Tasks and Risks of the OFR, First Principles ,OFR and Data, Data Collection, Standardizing Financial Contracts ,The Role of Standardized Contract Types .The Core Problem :Separation of Data and Algorithms in Natural Sciences , Separation of Data and Algorithms in the Financial Sector, The "Mechanical" Parts of Finance , The Subjective Parts of Finance , Combining the Mechanical and the Subjective .Appendix A: Exotic Products :Standardization in the Financial Sector, Standardization of the Outliers , The Boundary Between Standard and Nonstandard CTS. Appendix B: A Simple Data Model and Process: Introduction, Entity Relationship, The Contract Table and Validation, The Cash-flow Engine, The Mapping Process.

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Using Conceptual Models to Theorize about the Relationship between Records and Risk in the Global Financial Crisis: Records and Managerial Decisions, Information Problems and Growth, Information Problems and Financial Systems, Information Problems and Decision-Making in the Financial Industry, Analyzing the Financial Domain Using Conceptual Modelling, Conceptual Modelling as a Nontechnical Analysis Tool, Using Conceptual Modelling in the Analysis of Financial Transactions., Using Conceptual Modelling to Analyze the MBS Supply Chain, Analysis of the Conceptual Models. Financial Decision Process: Theory and Practice, The Experimental Methodology, Laboratory Experiments, Surveys/Financial Literacy, Visual Analytics for Financial Decision-Making, Decision Process and the Role of Visual Analytics, Information Processing and Search, Risk and Decision-Making, Research Challenges, Problems in Economics and Information Management, Moving Knowledge into Practice.

UNIT-IV A FUNCTIONAL FRAMEWORK FOR EVALUATING FINANCIAL VISUALIZATION 9 PRODUCTS

Introduction, Motivations for a Functional Evaluation Framework of VA in Finance, Developing a Functional Evaluation Framework of VA Tools, Methodology ,Evaluating a Sample Set of Visualizations and Interaction Techniques, Evaluating a Sample Set of VA Products and APIs, Limitations, The Functional Evaluation Framework of VA in Finance, Applied to Several Commercial and Research VA Toolkits, Case Study on the Process of Applying VA with Real-World Constraints in a Boutique Asset Management Firm, Conclusion and Future Work: Putting a Value on Visual Analytics, Case Study on the Process of Applying VA with Real-world Constraints in a Boutique Asset Management Firm, VA Solution.

UNIT-V PRESERVING DIGITAL RECORDS

9

Coping with Messiness and Fogginess in Financial Information Management: Material and Social Aspects of Representations in Proprietary Trading and Custodial Services : Introduction, Messiness and Fogginess, The

Invisibility of Financial Records and the Interplay Between Messiness and Fogginess, Proprietary Trading: Investing in Fogginess, Custodial Services: The Craft of Coping with Messiness and Fogginess. Preserving Digital Records: InterPARES Findings and Developments : Introduction ,Overview of the Three Phases, The Core Theoretical Foundation, Research Foci Research Design and Methodology ,Major Outcomes of InterPARES 1 and 2, InterPARES 3 TEAM Canada Findings and Products , Discussion of InterPARES Findings, Conclusion and Implications for Financial Data Management.

Total Contact Hours:45

G					
Co	Course Outcomes:				
On	completion of the course, students will be able to				
•	Have experience in the basics of Financial and Risk Analytics				
•	Apply programming knowledge of Operational Risk				
•	Manage financial decision making and Risk Management				
•	Gain how to evaluate Financial Visualization Products.				
•	Apply knowledge in implementing preserving digital records				

Tex	t Books:
	Financial Risk Analytics: A Term Structure Model Approach for Banking, Insurance, & Investment
1	Management Hardcover – Import, 1 Sep 996by Dennis G. Uyemura (Author), Donald R. van
	Deventer (Author)
	Financial Analysis and Risk Management Data Governance, Analytics and Life Cycle Management, Editor
2	Victoria Lemieux Library, Archival and Information Studies University of British Columbia Vancouver, BC
	Canada, Springer.
3	Credit Risk Analytics: Measurement Techniques, Applications, and Examples in SAS (Wiley and SAS Business
3	Series) 1st Edition, Kindle Edition by Bart Baesens (Author), Daniel Roesch (Author)
Ref	erence Books:
1	Financial Risk Management, Applications in Market, Credit, Asset and Liability Management and Firmwide
	Risk JIMMY SKOGLUND WEI CHEN, Wiley, Copyright © 2015 by SAS Institute.
2	Financial Analytics with RBuilding a Laptop Laboratory for Data Science by Mark J. Bennett, Dirk L. Hugen

РО	DO1	DO2	DO 2	DO4	DO5	DOC
со	PO1	PO2	PO3	PO4	PO5	PO6
C01	3	3	1	2	2	2
C02	3	3	1	2	3	1
C03	3	3	1	2	3	1
C04	3	3	1	2	3	2
C05	3	3	1	2	3	2
Average	3	3	1	2	2.8	1.6

Correlation levels 1, 2 or 3 are as defined below:

1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High)

No correlation : "-"

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19P33	MARKETING AND RETAIL ANALYTICS	PE	3	0	0	3
						1

Ob	Objectives:				
•	To understand the basics of Marketing				
٠	To understand and implement marketing Strategy				
٠	To understand the basics of product and services in marketing				
٠	To understand the retail analytics				
•	To apply predictive analytic techniques on retail and marketing				

UNIT-I	INTRODUCTION TO MARKETING	9
Statistical N	Aeasures, Understanding the marketplace and consumer needs, Designing a Customer Driven Mark	ceting-
Strategy, Bu	uilding Customer Relationships, Consumer Behaviour and Business Buyer Behaviour Analysis using	R
UNIT-II	MARKETING STRATEGY	9
Consumer I	Behaviour and marketing Strategy, Market Segmentation and Product Positioning, Market Segmen	tation,
Market Tar	geting, Target Market Strategies, Product Positioning and Differentiation, Choosing a Differentiation	on and
Positioning	Strategy,Logistic Regression for Market Basket Analysis	
UNIT-III	PRODUCT AND SERVICE	9
Products an	d services, product and service classifications, consumer products, industrial products, product and s	service
decisions, p	roduct and service attributes, product support services, services marketing - the nature and character	eristics
of a service,	Case Study-Panel Regression.	
UNIT-IV	RETAIL ANALYTICS	9
Customer A	Analytics Overview; Quantifying Customer Value. Using Stata for Basic Customer Analysis. Pred	licting
D		anothing
Response w	ith RFM Analysis, Statistics Review, Predicting Response with Logistic Regression, Predicting Response with Response with Logistic Regression, Predicting Response with Logistic Regression, Predicting Response with Re	
-	Networks, Predicting Response with Decision Trees.	
-		
with Neural UNIT-V	Networks, Predicting Response with Decision Trees.	sponse
with Neural UNIT-V The digital of	Networks, Predicting Response with Decision Trees. PREDICTIVE RETAIL ANALYTICS	sponse 9
with Neural UNIT-V The digital of	Networks, Predicting Response with Decision Trees. PREDICTIVE RETAIL ANALYTICS evolution of retail marketing, Digital natives, Constant connectivity Social Predictive modelling, Keeping track, Data availability, Efficiency optimization, Dependent Va	sponse 9

 Course Outcomes:

 On completion of the course, students will be able to

 •
 Understand the basics of Marketing

 •
 Implement marketing Strategy

 •
 Understand the basics of product and services in marketing

 •
 understand the retail analytics

 •
 Apply predictive analytic techniques on retail and marketing

Te	xt Books:
1	Kotler, P., Keller, K. L., Koshy, A., Jha, M. Marketing Management: A South Asian
1	Perspective. New Delhi: Pearson Education, 14th edn, 2013.
2	Mike Grigsby,Marketing Analytics:A practical guide to improving consumer insights using data techniques, Kogan Page 2 nd Edition.2018.
3	Rajan, S. Marketing Management. India: New Delhi: Tata McGraw-Hill Education. 4thedn,2005.

РО	DO1	DO1	DO3	PO4	PO5	DOC
СО	PO1	PO2	РОЗ	PO4	POS	PO6
C01	3	2	0	0	1	1
C02	3	3	0	0	1	1
C03	3	3	0	0	1	1
C04	2	2	0	0	2	2
C05	2	2	0	0	2	2
Average	2.6	2.4	0	0	1.4	1.4

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Theory course)	Category	L	Т	Р	С
DS19P34	SUPPLY CHAIN AND LOGISTICS ANALYTICS	PE	3	0	0	3

Ob	Objectives:					
•	To understand the importance of Supply Chain Analytics and Network Planning Design of Logistics Network					
•	To analyze the level of uncertainty associated with the supply of products and services to targeted customer segments					
٠	To explicate applications of Analytics in a Supply Chain.					
•	To explore the technical standards and Business processes integrated with Supply Chain					

UNIT-I INTRODUCTION TO SUPPLY CHAIN

Supply chain Management-Introduction-Evolution-Analytics-Supply chain planning- Different views of Supply chain-Key issues-Supply chain strategy-Supply Chain Drivers-Strategic fit in Supply chain-Demand Forecasting in Supply chain-Case Study: Meditech Surgical.

UNIT-II NETWORK PLANNING AND VALUE OF INFORMATION

Network Design-parameters and requirements of network-Inventory positioning and Logistics Co-ordination-Case Study: Bis Corporation-Supply Contracts- Strategic Components-Contracts -Make to Stock-Asymmetric In-formation-Non Strategic Components-Bull whip effect-Sharing and Incentives-Effective Forecast-Coordination of Systems-Lead-Time Reduction-Case study: Reebok NFL Replica Jerseys.

UNIT-III DISTRIBUTION STRATEGY

Push, Pull, Push-Pull Systems-Impact of Lead Time-Demand driven Strategies-Impact of Internet on Supply Chain Strategies – Case Study: Great Inventory Correction-Distribution Strategies –Direct Shipment-Intermediate Storage Points-Transshipment-Selection of Strategy-Case Study: Amazon.com's European Distribution Strategy-Third Party Logistics-Retailer-Supplier Partnerships-Distributor Integrator.

UNIT-IV STRATEGIC ALLIANCES

Outsourcing-Benefits-Risks-Procurement Strategies-E-Procurement-Case study: Zara-Solectron-Forces-Types-Risk management-Issues in Supply Chain management-Regional Differences-Case Study: Walmart-Logistics-Design-Supplier Integration-Customer Value-Smart Pricing-Case Study: Made to Measure-Start Bucks Economics.

UNIT-V IT ENABLEMENT OF SUPPLY CHAIN

Goals-Decision Support System-IT for Supply Chain Excellence-Sales and Operation Planning-Supply Chain with Information Technology-Case Study: Supply Chain Whirl-Technology Standard-IT standard -IT Infrastructure-SOA-RFID-Case Study: Amazon story of Dabbawallas.

Total Contact Hours:45

0

9

0

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Course Outcomes:

	in se Outcomes.						
On	On completion of the course, students will be able to						
•	Identify and Analyze Business Models, Business Strategies and, corresponding Competitive Advantage.						
•	Formulate and implement Best Practices and Strategies for Design and Distribution						
•	Plan the effective Logistics operations for optimum utilization of resources						
•	Configure a supply chain for an organization from a global perspective that accounts for commercial, social and legal implications.						
•	Discuss the impacts of geo-political and technological trends/developments on the value chain.						

Ref	Reference Books(s) / Web links:							
1	D. Simchi-Levi, P. Kaminsky, E. Simchi-Levi, and Ravi Shankar, Designing and Managing the Supply Chain concepts, Strategies and Case studies, Third Edition, Tata McGraw Hill, New Delhi, 2008.							
2	Supply chain management by Sunil Chopra, and Peter Meindl, Pearson Jeremy F. Shapiro. Modeling the Supply Chain. Duxbury Thomson Learning							
3	Supply Chain and Logistics Management Made Easy, Paul A. Myerson, Person Education, Inc., 2015.							
4	Chadwick, T. and Rajagopal, S. (1995): Strategic Supply Management: An Implementation Toolkit, Butterworth-							
4	Heinemann, Oxford, UK.							
5	Jeremy F. Shapiro. Modeling the Supply Chain. Duxbury Thomson Learning, 2001							
6	https://nptel.ac.in/courses/110107074/							
7	Journal: Supply Chain Management: An International Journal (Available on www.emaraldinsight.com)							

РО	DOI	DOA	DOA	DO 4	D O 5	DOC
СО	PO1	PO2	PO3	PO4	PO5	PO6
C01	3	2	-	-	2	-
C02	3	3	2	-	-	2
C03	2	1	1	-	1	1
C04	2	2	2	-	-	2
C05	3	2	-	-	-	2
Average	2.6	2	1.6	0	1.5	1.75

Correlation levels 1, 2 or 3 are as defined below:

Subject Code	Subject Name (Theory course)		L	Τ	Р	C
CP19O31	BUSINESS ANALYTICS	OE	3	0	0	3

Objectives	S
•	To understand the role of business analytics within an organization.
•	To analyze data using statistical and data mining techniques.
•	To gain an understanding of how managers use business analytics to formulate and solve business
	problems.
•	To become familiar with processes needed to develop, report, and analyze business data.
•	To use decision-making tools/Operations research techniques.

UNIT-I INTRODUCTION

Business analytics: Overview of Business analytics- Scope of Business analytics- Business Analytics Process-Relationship of Business Analytics Process and organization – competitive advantages of Business Analytics-Statistical Tools: Statistical Notation- Descriptive Statistical methods- Review of probability distribution and data modelling- sampling and estimation methods overview.

UNIT-II **REGRESSION AND VISUALIZATION**

Trendiness and Regression Analysis: Modelling Relationships and Trends in Data- simple Linear Regression. - Important Resources- Business Analytics Personnel- Data and models for Business analytics- problem solving-

Visualizing and Exploring Data-Business Analytics Technology.

UNIT-III ANALYTICAL MODELS

Organization Structures of Business analytics-Team management-Management Issues- Designing Information Policy- Outsourcing- Ensuring Data Quality- Measuring contribution of Business analytics- Managing Changes. Descriptive Analytics- predictive analytics- predicative Modelling- Predictive analytics analysis-Data Mining and Methodologies- Prescriptive analytics and its step in the business analytics Process-Prescriptive Modelling-nonlinear Optimization. 9

UNIT-IV FORECASTING TECHNIQUES

Qualitative and Judgmental Forecasting- Statistical Forecasting Models- Forecasting Models for Stationary Time Series- Forecasting Models for Time Series with a Linear Trend- Forecasting Time Series with Seasonality-Regression Forecasting with Casual Variables- Selecting Appropriate Forecasting Models- Monte Carlo Simulation and Risk Analysis: Monte Carle Simulation Using Analytic Solver Platform– New-Product Development Model–Newsvendor Model– Overbooking Model– Cash Budget Model.

UNIT-V DECISION ANALYSIS

Decision Analysis: Formulating Decision Problems- Decision Strategies with the without Outcome Probabilities- Decision Trees- The Value of Information, Utility and Decision Making- Recent Trends in: Embedded and collaborative business intelligence- Visual data recovery- Data Storytelling and Data ournalism.

Total Contact Hours 45 :

9

9

9

9

Course Outcomes:

Upon completion of the course, the students will be able to Demonstrate knowledge of data analytics. • Think critically in making decisions based on data and deep analytics. • Use technical skills in predicative and prescriptive modeling. ٠ Translate data into clear, actionable insights. • Make decisions using various tools. Reference Book(s): Marc J. Schniederjans, Dara G. Schniederjans, Christopher M. Starkey, "Business analytics Principles, 1 Concepts,

and Applications", Pearson FT Press, 2014.

PO/CO	PO1	PO2	PO3	PO4	PO5	PO6
C01	2	3	2	2	2	3
C02	2	2	1	2	3	2
C03	3	2	3	3	2	2
C04	2	3	3	2	3	2
C05	2	3	3	2	3	3
Average	2.2	2.6	2.4	2.2	2.6	2.4

Correlation levels 1, 2 or 3 are as defined below:

1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High)

No correlation : "-"