Resolution No.: AC/II(20-21).2.RPS6

S.P. Mandali's RAMNARAIN RUIA AUTONOMOUS COLLEGE



Syllabus for: M.Sc. Computer Science

Program: M.Sc.

Course Code: Computer Science (RPSCS)

(Choice Based Credit System (CBCS) with effect from the academic year 2020-21)

MSC Computer Science SYLLABUS

CREDIT BASED SYSTEM AND GRADING SYSTEM

The ACADEMIC YEAR 2020 - 2021

Semester-I: Theory courses

Each of these theory courses is of four credits each and is expected to complete in 60 hours. The details are shown in the following table.

SEMESTER - I (THEORY)			
COURSE CODE	COURSE TITLE	CREDITS	LECTURES/WEEK
RPSC 101	Analysis of Algorithm	04	04
RPSCS102	Advanced-Data Modelling	04	04
RPSCS103	Natural Language Processing	04	04
RPSCS104	Blockchain	04	04

SEMESTER - I (PRACTICALS)				
COURSE CODE	COURSE TITLE	CREDITS	PRACTICALS /WEEK	
RPSCSP101	Analysis of Algorithm	02	04	
RPSCSP102	Advanced Data Modelling	02	04	
RPSCSP103	Natural Language Processing	02	04	
RPSCSP104	BlockChain	02	04	

Informationster-II: Theory courses

Each of the courses is expected to complete in 60 hours. The details are given in the following table.

SEMESTER – II (THEORY)				
COURSE CODE	COURSE TITLE	CREDITS	LECTURES/W EEK	
RPSCS201	Machine Learning	04	04	
RPSCS202	Cyber & information Security	04	04	
RPSCS203	Wireless and Mobile Networks	04	04	
RPSCS204	Advanced Mining Techniques	04	04	

SEMESTER - I	II (PRACTICALS)	40	9
COURSE CODE	COURSE TITLE	CREDITS	LECTURES/W EEK
RPSCSP201	Machine Learning	02	04
RPSCSP202	Cyber & information Security	02	04
RPSCSP203	Wireless and Mobile Networks	02	04
RPSCSP204	Advanced Mining Techniques	02	04

Semester-III: Theory courses

SEMESTER – III (THEORY)				
COURSE TITLE CREDITS LECTURES/WEEL				
RPSCS301	Social Network Analysis	04	04	
RPSCS302	Internet of Things	04	04	
RPSC 303	Big Data using Hadoop and R	04	04	

SEMESTER - III (PRACTICALS)			
COURSE CODE	COURSE TITLE	CREDITS	PRACTICALS /WEEK

RPSCSP301	Social Network Analysis	04	04
RPSCSP302	Internet of Things	04	04
RPSCSP303	Big Data using Hadoop and R	04	04
RPSCSP304	Project	06	

SEMESTER – IV (Theory)			
COURSE CODE	COURSE TITLE	CREDITS	LECTURES/WEEK
RPSCS401	Simulation and Modelling	04	04

SEMESTER – IV (Practical)				
COURSE CODE	COURSE TITLE	CREDITS	LECTURES/WEEK	
RPSCSP401	Simulation and Modelling	02	04	
RPSCSP402	Internship	18		

Project: The student is expected to develop an industry or research-based major project in group max size of three or individually on the topic of his/her choice after getting approval from the respective project guide assigned to the student/s by the department. The student is expected to show regular updates and reviews to the department as scheduled from time to time.

SEMESTER I- THEORY Detailed Syllabus

COURSE CODE	COURSE NAME	CREDITS	LECTURE/ WEEK
RPSCS101	RPSCS101: Analysis of Algorithms	4	4

Course Objective:

- Understanding Algorithmic complexity and analyzing the same
- Developing an understanding of various techniques and methods to design algorithms

Expected Learning Outcome:

Skill to make the algorithm and solve real-world problems

UNITS	COURSE CONTENTS	NO. OF LECTURES
I	Introduction The Role of Algorithms in Computing, Algorithms as a technology, Insertion sort, Analyzing algorithms, Designing algorithms, Growth of Functions, Asymptotic notation, Standard notations and common functions, Divide-and-Conquer, The maximum-subarray problem, Strassen's algorithm for matrix multiplication, The substitution method for solving recurrences, The recursion-tree method for solving recurrences, The master method for solving recurrences, Proof of the master theorem. Probabilistic Analysis and Randomized Algorithms: The hiring problem, Indicator random variables, Randomized algorithms, Probabilistic analysis and further uses of indicator random variables	
II	Sorting and Order Statistics: Introduction, Heapsort, Heaps, Maintaining the heap property, Building a heap, The heapsort algorithm, Priority queues, Quicksort, Description of quicksort, Performance of quicksort, A randomized version of quicksort, Analysis of quicksort, Sorting in Linear Time, Lower bounds for sorting, Counting sort, Radix sort, Bucket sort, Medians and Order Statistics, Minimum and maximum, Selection in expected linear time, Selection in worst-case linear time.	
III	Dynamic Programming, Rod cutting, Matrix-chain multiplication, , Elements of dynamic programming, Longest common subsequence, Optimal binary search trees, Greedy Algorithms, An activity-selection problem, Elements of the greedy strategy, Huffman codes, Matroids and greedy methods, A task-scheduling problem as a matroid, Amortized Analysis, Aggregate analysis, The accounting method, The potential method, Dynamic tables	
IV Text book:	String Matching: The naive string-matching algorithm, The Rabin-Karp algorithm, String matching with finite automata, The Knuth-Morris-Pratt algorithm NP-Completeness: Polynomial-time, Polynomial-time verification, NP-completeness and reducibility, NP-completeness proofs, NP-complete problems, Approximation Algorithms, The vertex-cover problem, The travelling-salesman problem, The set-covering problem, Randomization and linear programming, The subset-sum problem	

 Introduction to Algorithms Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, Clifford Stein

References:

- 1. Analysis and Design of Algorithms: A Beginner's Approach by Rajesh K. Shukla
- 2. Design and Analysis of Algorithms: A Contemporary Perspective by Sandeep Sen

COURSE CODE	COURSE NAME	CREDITS	LECTURE/ WEEK
RPSCSP101	RPSCSP101: PRACTICAL OF Analysis of Algorithms	2	4
1	Demonstrate Strassen Matrix Multiplication.		
2	Solve the Hiring Problem with an appropriate algorithm		
3	Demonstrate various sorting algorithm and compute the efficiency of each		
4	Demonstrate dynamic programming with Rod Cutting problem.		
5.	Demonstrate the Long Subsequence problem and its solution using recursion.		
6	Demonstrate the Huffman Codes and its application.		
7.	Implement The Knuth-Morris-Pratt algorithm		
8.	Demonstrate the travelling Sales man's problem		

COURSE CODE	COURSE NAME	CREDITS	LECTURE/ WEEK
RPSCS102	RPSCS102: Advanced Data Modelling	4	4

Course Objective:

- To acquire knowledge on a variety of advanced data modelling techniques.
- To have a comparative study of different NoSql products and Applications

• Design and implement case studies based on different data models

Expected Learning Outcome:

- Demonstrate competency in designing spatial and semi-structured databases
- Expertise in selecting a particular NoSQL database for specific use cases.
- Implementing different technologies of handling spatial, semistructured and NoSQL data.

UNITS	COURSE CONTENTS	NO. OF LECTURES
I	Spatial data modelling: spatial data types, Raster, vector representation, topological spatial relationships, indexing, Geographical information systems: Data management, processing, referencing and analysis, visualization	15 L
II	Semi-structured data modelling: Semi-structured data model, Hierarchical data model, DTD, XML documents and schema, XML documents and databases, XML querying	15 L
III	Nosql: Features, Managing different data types, consistency methods, Distributed scenario, Partitioning, Query model, Storage layout, Enterprise application Evaluating Nosql: Technical, Business, issues	
IV	NoSQL data stores: features and use cases of key-value, Big Table, document database, hybrid NoSQL, search engines.	15 L

Text book:

- 1. NoSQL for Dummies by Adam Fowler published by Willey
- 2. NoSQL and SQL Data modelling By Ted Hills
- 3. Fundamentals of database systems Navathe
- 4. Principles of GIS, otto Huisman, Rolf A

COURSE CODE	COURSE NAME	CREDITS	LECTURE/ WEEK
RPSCSP102	RPSCSP102: Advanced Data Modelling	2	4
	Real-world Case studies based on the following single categrombination of multiple databases 1. Semi-structured data 2. Spatial data 3. GIS 4. Nosql: Columnar data store 5. NoSQL: Key-value datastore 6. NoSQL: Graph datastore 7. NoSQL: Document datastore 8. NoSQL: Search engine	ory of databa	ases or a

COURSE CODE	COURSE NAME	CREDITS	LECTURE /WEEK
RPSCS103	RPSCS103: Natural Language Processing	4	4

Course Objective:

- To learn how to apply basic algorithms in this field.
- To tag a given text with basic Language features
- To design and implement applications based on natural language processing
- To implement a rule-based system to tackle morphology/syntax of a language
- To design an innovative application system that uses NLP components

Expected Learning Outcome:

- To get a deeper understanding of the field of natural language processing.
- Have a sense of the capabilities and limitations of natural language technologies.
- Be able to model linguistic phenomena with formal grammars.
- Be able to design, implement and test algorithms for NLP problems
- To compare and contrast the use of different statistical approaches for different types of NLP applications.
- Be able to apply NLP techniques to design real-world NLP applications

UNITS	COURSE CONTENTS	
I	Introduction: Knowledge In Speech And Language processing, Ambiguity, History of NLP. Words Analysis: Regular expression, Regular Languages and FSA, Words & Transducers, Morphology, Words & sentence Tokenization, Human Morphological processing, N-Grams, Evaluating N-grams, Parts of speech tagging (POST), Hidden Markov Model (HMM).	
II	Speech: Phonetics, Speech sound & phonetic transcription, Articulatory phonetics, Speech Synthesis, Automatic Speech recognition, Computational phonology.	
III	Syntax: Formal Grammars of English, Parsing with context-free grammars, Statistical Parsing, Language and complexity, Features and unification	15 L
IV	Semantics & applications: Representing Meaning, Computational semantic, lexical-semantic, Computational lexical-semantic, Information Extraction, Machine translation.	

Text book:

- 1. Speech and Language Processing By Daniel Jurafsky, James H. Martin 2nd Edition, Prentice-Hall, 2008/2009.
- 2. Foundations of Statistical Natural Language Processing By Christopher D. Manning and Hinrich Schutze -The MIT Press (1999), Cambridge, Massachusetts, London, England.
- 3. Natural Language Processing with Python" Analyzing Text with Natural Language Toolkit. -By Steven Bird, Ewan Klein, Edward Loper, O'Reilly Media

- 1. Natural Language Processing: A Paninian Perspective By Akshar Bharati, Vineet Chaitanya, Rajeev Sangal.
- 2. Steven Bird, Ewan Klein and Edward Loper, Natural Language Processing with Python, First Edition, OReilly Media, 2009.

COURSE CODE	COURSE NAME	CREDITS	LECTURE/W EEK
RPSCSP103	RPSCSP103: PRACTICAL OF Natural Language Processing	2	4
1.	Preprocessing of text: Word Analysis (Tokenization, Filtration, Script Validation)		c.A
2.	Preprocessing of text: Word Generation (word frequency, Stop Word Removal, Stemming)	COL	
3.	Morphological Analysis		
4.	N-gram model		
5.	POS tagging: HMM		
6.	POS tagging: Viterbi Decoding		
7.	Building POS Tagger		
8.	Chunking		
9.	Building Chunker / Name Entity Extraction		
10.	Case Study based on Application		

COURSE CODE	COURSE NAME	CREDITS	LECTURE/ WEEK
RPSCS104	RPSCS104: Blockchain	4	4

Course Objective:

• To understand blockchain technology as a security mechanism

Expected Learning Outcome:

- Understand blockchain types and implementation
- Construct and deploy Smart contracts using Ethereum
- Able to implement decentralized applications using blockchain in real-time application.

UNITS	COURSE CONTENTS	NO. OF
UNITS	COURSE CONTENTS	LECTURES

-			
	I	Introduction to blockchain: Web3 - The Decentralized Web Distributed system, History of blockchain, Need of blockchain Satoshi Nakamoto's Blockchain Breakthrough, Basics of blockchain, Block structure, Features, Types of blockchain, Application of blockchain, Consensus in blockchain, Benefits, limitation and challenges of blockchain, Decentralization: Methods of decentralization, Roles to decentralization, Smart contracts, Decentralized organization Public blockchain: Basics of Bitcoin, transactions, blockchain, Mining, Walle types, Bitcoin network, Bitcoin limitations, Ethereum blockchain, elements of Ethereum blockchain, Block structure, Mining and consensus, Construction of smar contract by solidity, Alternative coins, Bitcoins,namecoin, Peercoin Private blockchain: Hyperledger GreenHouse, Introduction to Hyledger Fabric Features, Architecture, Requirements, Components of fabric, Working of fabric Hyperledger SAWTOOTH, Hyperledger BESU Hyperledger INDY, Hyperlege BURROW	
	II		
	III		
	IV	Blockchain Use cases: Business Use Cases, Technology Use Cases, Legal and Governance Use Cases, Private blockchain Use cases	15 L

Text book:

- 1. Mastering Blockchain, Imran Bashir,packt
- 2. A Practical Guide to Developing Business, Law, and Technology Solutions, Joseph J. Bambara Paul R. Allen

- 1. https://bitcoin.org/bitcoin.pdf
- 2. Mastering Bitcoin by Andreas M. Antonopoulos, O'Reilly
- 3. Mastering Etherium, Andreas M. Antonopoulos, O'Reilly
- 4. <u>€https://solidity.readthedocs.io/en/v0.6.2/</u> (for solidity documentation
- 5. https://www.hyperledger.org/
- 6. https://www.leewayhertz.com/blockchain-platforms-for-top-blockchain-companies

COURSE CODE	COURSE NAME	CREDITS	LECTURE/WE EK
RPSCSP104	PRACTICAL OF RPSCSP104 Block chain	2	4
1	Double Linked list Implementation		

2	Hashing algorithms		
3	Cryptography Fundamentals		A.C.
4	Simple blockchain implementation		Tere
5	Blockchain with MongoDB	Co	
6	Ethereum Smart contract development		
7	Ethereum Dapp		
8	Hyper ledger Fabric		
9	Blockchain Case Study		

3	Dicertain Case Stady		
	SEMESTER II- THEORY Detailed Syllabus		
COURSE CODE	COURSE NAME	CREDITS	LECTURE /WEEK
RPSCS201	RPSCS201: Machine Learning	4	4
Course Obje	ctive: standing the Fundamental of Machine Learning Algorithms	1	

• Developing skills to make and device ML strategies to solve real-life problems

Expected Learning Outcome:

Making use of old and new techniques of ML to solve problems

I	Artificial Neural Networks The Artificial Neuron, Supervised Learning Neural Networks, Unsupervised Learning Neural Networks, Radial Basis Function Networks, Reinforcement Learning, Performance Issues.	
II	Evolutionary Computation Introduction to Evolutionary Computation, Genetic Algorithms, Genetic Programming, Evolutionary Programming, Evolution Strategies, Differential Evolution, Cultural Algorithms, Co-evolution.Computational Swarm Intelligence: Particle Swarm Optimization(PSO) - Basic Particle Swarm Optimization, Social Network Structures, Basic Variations and parameters, Single-Solution PSO.	
III	Advanced Topics and applications. Ant Algorithms- Ant Colony Optimization Meta- Heuristic, Cemetery Organization and Brood Care, Division of Labor, Advanced Topics and applications. Artificial Immune systems, Fuzzy Systems and Rough Set Natural Immune System, Artificial Immune Models, Fuzzy Sets, Fuzzy Logic and Reasoning, Fuzzy Controllers, Rough Sets.	l t
IV	Deep Learning: Deep forward Networks, Regularization of Deep Learning, Optimization of Train Deep Models, Convolution Networks, Sequence Modeling: Recurrent and Recursive Nets	

Text Book:

- 1. Computational Intelligence- An Introduction (Second Edition): Andries P.Engelbrecht, John Wiley & Sons Publications (2007).
- 2. Deep Learning Ian Goodfellow, Yoshua Bengio, Aaron Courville MIT Press

- 1. Computational Intelligence And Feature Selection: Rough And FuzzyApproaches, Richard Jensen Qiang Shen, IEEE Press Series On ComputationalIntelligence, A John Wiley & Sons, Inc., Publication, 2008.
- 2. Computational Intelligence And Pattern Analysis In Biological Informatics, (Editors). Ujjwal Maulik, Sanghamitra Bandyopadhyay, Jason T. L.Wang, JohnWiley & Sons, Inc, 2010.
- 3. Neural Networks for Applied Sciences and Engineering: From Fundamentals to Complex Pattern Recognition 1st Edition, Sandhya Samarasinghe, Auerbach Publications, 2006.
- 4. Introduction to Evolutionary Computing (Natural Computing Series) 2nd ed, A.E. Eiben, James E Smith, Springer; 2015.
- 5. Swarm Intelligence, 1st Edition, Russell C. Eberhart, Yuhui Shi, James Kennedy, Morgan Kaufmann, 2001
- 6. Artificial Immune System: Applications in Computer Security, Ying Tan, Wiley-IEEE Computer

Society, 2016.

7. Computational Intelligence and Feature Selection: Rough and Fuzzy Approaches 1st Edition, Richard Jensen, Qiang Shen, Wiley-IEEE Press, 2008

COURSE CODE	COURSE NAME	CREDITS	LECTURE/ WEEK
RPSCSP201	RPSCSP201: PRACTICAL for Machine Learning	2	4
1.	Implement a feed-forward neural network for the given data.		
2.	Implement a Self Organizing Map neural network.		
3.	Implement a Radial Basis Function neural network with gradient descent.		
4.	Implement a basic genetic algorithm with selection, mutation and crossover as genetic operators.		
5.	Implement an evolution strategy algorithm.		
6.	Implement a general differential evolution algorithm.		
7.	Implement a simple Ant colony optimization algorithm.		
8.	Demonstrate Convolution networks using a deep learning model.		

COURSE CODE	COURSE NAME	CREDITS	LECTURE/ WEEK
RPSCS2 02	Cyber and Information security	4	4

Course Objective:

• To understand security mechanisms to secure data

Expected Learning Outcome:

- Secure data on different platforms
- Identify possible attacks on clouds and its countermeasures

UNITS	COURSE CONTENTS	NO. OF LECTURES
I	Server Security Server vulnerabilities, Threats and environment, Basic server security steps, Server security principle, Server security planning, Securing server operating system, Windows security, Unix security, Securing server database system, Securing server software, Maintaining the security of the server, virtual machines, Secure application design, Developing secure software	15 L
II	Network Security Secure Network Design, Network device security, Firewalls, Virtual private networks, Wireless network security, IDPS, VOIP and PBX security, SET, Detection and response to attacks, Risk analysis and management, Network security case scenario	15L
III	Cloud Security Cloud Computing Software Security Fundamentals, Cloud Computing Risk Issues, Cloud Computing SecurityChallenges, Cloud Computing SecurityArchitecture, Cloud Computing Life Cycle Issues, cloud security case scenario	15 L
IV	IOT Security: Threats AND ATTACKS, Computational Security for the IoT, Privacy-Preserving Time, Series Data Aggregation for the Internet of Things, Trust and authentication, Secure Path Generation, Scheme for Real-Time Green Internet of Things, Security Protocols for IoT Access Networks	15 L

Text book:

- 1. The Complete Reference: Information Security, Mark Rhodes-Ousley, McGraw-Hill 2nd Edition, 2013
- 2. Cloud Security A Comprehensive Guide to Secure Cloud Computing, Ronald L.Krutz, Russell Dean Vines, Wiley Publishing, Inc., 2010.
- 3. Security and Privacy in Internet of Things (IoT), Fei Hu, CRC press

- 1. Guide to General Server Security, Recommendations of the National Institute Of Standards and Technology, Karen Scarfone Wayne Jansen Miles Tracy
- 2. Cloud Security and Privacy, Tim mather, Subra Kumaraswamy
- 3. Securing Cloud and Mobility A Practitioner's Guide, Ian Lim. Coleen CoolidgePaul Hourani
- 4. Security in Computing 4th edition, Charles P. Pfleeger, Charles P.Pfleeger, Shari Lawrence Pfleeger, Prentice Hall; 4th edition (2006)

COURSE CODE	COURSE NAME	CREDITS	LECTURE/WEEK
RPSCSP2 02	PRACTICAL OF RPSCSP202 Cyber & Information Security	2	4
1.	Secure Communication		Corre
2.	Secure programming implementation		
3.	Server Security	Mo	
4.	Database Security		
5.	cloud database security		
6.	cloud server Security		
7.	Using Docker for deployment to Secure Appliction acode		
8.	Network monitoring tool		

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COURSE CODE	COURSE NAME		CREDITS	LECTURE /WEEK
RPSCS203	RPSCS203: Wireless and Mobile	e Networks	4	4
Course Object	ive:			

- Understanding Wireless Networking Systems
- Working with Mobile Technologies and innovations in the same.

Expected Learning Outcome:

• Using wireless networking for developing applications and projects

UNITS	COURSE CONTENTS	NO. OF LECTURE S
I	Introduction to Telephone Systems, Telephones, Control Functions telephone Traffic, Switching, Wireless information Networks, Information Transmission, State Diagram of Telephone Network, Modems, Mobile Communication, Need, Requirements, History, Properties of Wireless Medium. Introduction to Cellular Mobile Communication, Cellular Structure Frequency Reuse, System Architecture, Traffic and Switching Techniques Mobile Communication Standards: Generation of Wireless networks Standard Organizations,	
II	Global Systems for Mobile Communication (GSM), GSM Architecture, AMPS, DAMPS, Cordless Telephony, PACS, Third Generation Wireless Standards. Mobility Management: Handoff Techniques, Handoff Detection and Assignment, Types of Handoff, Radio Link transfer, Roaming Management. Frequency Management, Cellular System Spectrum, Adaptive Channel allocation, Frequency Division, Spectrum Utilization, channel reservation for handoff calls, control channels, channel assignment methods, cell splitting. Mobile Computing, Classification of mobile data networks, Cellular digital packet data (CDPD) Systems. Satellite in Mobile Communication. Global Mobile Communication, Interferences in cellular communication, Mobile internet	
III	Wireless Security: Wireless Threats, Authentication and access control, secrecy in communication, Security arrangements in CDMA, Security of wireless data networks, Wireless Local Loop Architecture: Components in WLL, Problems in WLL, Modern Wireless Loop, LDMS, Wireless Application Protocol: Properties of WAP, Bearer Services, WAP Component Integration, WAP Client Support Networks.	
	WCDMA and Fibre optic microcellular Mobile communication: System Description, Layout of Optical Fibre and Microcellular Communication System, Adhoc Networks and Bluetooth Technologies: MANET and Technical Factors affecting ad hoc networks, Bluetooth technology Intelligent Mobile Communication: Types of Intelligent Cell, Power delivery Intelligent Cells, Processing Gain intelligent cells, 4G: User-Controlled services, Reconfigurable Technology, Vision of 4G, 4G Mobile System Convergences.	

Text book:

"Wireless and Mobile Communication" by T.G. Nakkeeran, R. Palanivelu

- 1. Wireless Communications: Principles and Practice, 2e Paperback 2010 by Rappaport
- 2. Wireless Communications & Networks, 2e Paperback 2009 by Stallings

3. Mobile Communications, 2e Paperback – 2008 by Schiller

COURSE CODE	COURSE NAME	CREDIT S	LECTURE/ WEEK
RPSCSP203	RPSCSP203: Practical of Wireless and Mobile Networks	2	4
1.	Demonstrate Frequency distribution in Cellular Network		
2	Demonstrate and over in Adhoc networks		
3	Demonstrate satellite communication with earth stations		
4	Illustrate the wireless local loops.		
5.	Demonstrate the LDM systems		
6.	Develop a Bluetooth ad hoc network		
7.	Demonstrate the working of an intelligent cell		
8	Demonstrate MANET		

COURSE CODE	COURSE NAME	CREDITS	LECTURE/W EEK
RPSCS204	RPSCS204: Advanced Mining Techniques	4	4

Course Objective:

- Understanding Data Mining algorithms and strategies
- Developing skills to solve real-life problems in data management

Expected Learning Outcome:

• Making use of old and new techniques of data mining to solve problems

UNITS	COURSE CONTENTS	NO. OF LECTURE S
I	What is data Statistical Modeling?, Machine Learning, Computational Approaches to Modeling, Summarization, Feature Extraction Statistical Limits on Data Mining. MapReduce and the New Software Stack: Distributed File Systems, MapReduce, Algorithms Using MapReduce, Extensions to MapReduce, The Communication Cost Model, Complexity Theory for MapReduce	
II	Finding Similar Items: Applications of Near-Neighbor Search, Shingling of Documents, Similarity-Preserving Summaries of Sets, Locality-Sensitive Hashing for Documents, Distance Measures, The Theory of Locality-Sensitive Functions, LSH Families for Other Distance Measures, Applications of Locality-Sensitive Hashing, Methods for High Degrees of Similarity. Mining Data Streams: The Stream Data Model, Sampling Data in a Stream, Filtering Streams, Counting Distinct Elements in a Stream, Estimating Moments, Counting Ones in a Window, Decaying Windows	
III	Link Analysis: PageRank, Efficient Computation of PageRank, Topic-Sensitive PageRank, Link Spam, Hubs and Authorities. Frequent Itemset: The Market-Basket Model, Market Baskets and the A-Priori Algorithm Handling Larger Datasets in Main Memory, Limited-Pass Algorithms, Counting Frequent Items in a Stream.	
IV	Clustering: Introduction to Clustering Techniques, Hierarchical Clustering, K-means Algorithms, The CURE Algorithm, Clustering in Non-Euclidean Spaces, Clustering for Streams and Parallelism. Recommendation Systems: A Model for Recommendation Systems, Content-Based Recommendations, Collaborative Filtering, Dimensionality Reduction.	

Text book:

1. Mining of Massive Datasets by Jure Leskovec Stanford Univ., Anand Rajaraman Milliway Labs, Jeffrey D. Ullman Stanford Univ.

References:

1) Data mining concepts and techniques by Jiawei Han, Micheline Kamber, Jian Pei 3rd Edition.

Semester III Detailed Syllabus

COURSE CODE	COURSE NAME	CREDITS	LECTURE /WEEK
RPSCS301	RPSCS301: Social Network Analysis	4	4

Learning Objective:

- understand the behaviour of the users in the social network
- Predict the possible next outcome of the social network

Learning Outcome:

- Work on the internal components of the social network
- Model and visualize the social network
- Apply social network in real-time applications

UNITS	COURSE CONTENTS	NO. OF LECTURE S
I	Introduction to social network analysis (SNA)	15 L
	Introduction to networks and relations- analyzing relationships to	
	understand people and groups, binary and valued relationships,	
	symmetric and asymmetric relationships, multimode relationships, social	
	networks vs. link analysis, ego-centric and socio-centric density.	
	Visualization and applications of social networks	
	Using graph theory for social networks analysis- adjacency matrices,	
	edge-lists, adjacency lists, graph traversals and distances, depth-first	
	traversal, breadth-first traversal paths and walks, Dijkstra's algorithm,	
	graph distance and graph diameter, Hybrid representation, Applications	
	- networks, community welfare, collaboration networks, Co-citation	
	networks.	45.1
II	Networks, Centrality and centralization in SNA	15 L
,0	Understanding networks- density, reachability, connectivity, reciprocity, group-external and group-internal ties in networks, ego networks,	
	extracting and visualizing ego networks, structural holes, Centrality	
	degree of centrality, closeness and betweenness centrality, local and	
A 37"	global centrality, centralization and graph centres, the notion of	
	importance within the network, Google Page Rank algorithm, Analyzing	
	network structure-bottom-up approaches using cliques, N-cliques, N-clans, K-plexes, K-cores, blocks and cut-points, lambda sets and	
	bridges & factions.	
III	Measures of similarity and structural equivalence in SNA	15 L
	Approaches to network positions and social roles- defining equivalence	
	similarity, structural equivalence, automorphic equivalence, finding	

equivalence sets, brute force and Tabu search, regular equivalence, the equivalence of distances: Maxim, regular equivalence, Measuring similarity/dissimilarity- valued relations, Pearson correlations covariance and cross-products, Euclidean, Manhattan, and squared distances, binary relations, Jaccard, Hamming distance	
Two-mode networks for SNA	15 L
Understanding mode networks- Bi-partite data structures, visualizing	
two-mode data, Singular value decomposition (SVD) analysis, two-mode	
Predicting Human Behavior and Privacy Issues	
Understanding and predicting human behaviour for social communities,	
User data Management, Enabling human experience, Privacy on social networks	
	equivalence of distances: Maxim, regular equivalence, Measuring similarity/dissimilarity- valued relations, Pearson correlations covariance and cross-products, Euclidean, Manhattan, and squared distances, binary relations, Jaccard, Hamming distance Two-mode networks for SNA Understanding mode networks- Bi-partite data structures, visualizing two-mode data, Singular value decomposition (SVD) analysis, two-mode factor analysis. Predicting Human Behavior and Privacy Issues Understanding and predicting human behaviour for social communities, User data Management, Enabling human experience, Privacy on social

- 1. Introduction to Social Network Methods: Robert A. Hanneman, Mark
- 2. Riddle, University of California, 2005 [Published in digital form and available athttp://faculty.ucr.edu/~hanneman/nettext/index.html].
- 3. Social Network Analysis for Startups- Finding connections on the social web: MaksimTsvetovat, Alexander Kouznetsov, O'Reilly Media, 2011.
- 4. Social Network Analysis- 3rd edition, John Scott, SAGE publications, 2012.
- 5. Ajith Abraham, Aboul Ella Hassanien, Václav Snášel, —Computational Social Network Analysis: Trends, Tools and Research Advancesll, Springer, 2012
- 6. Charu C. Aggarwal, —Social Network Data AnalyticsII, Springer; 2014

COURSE CODE	COURSE NAME	CREDITS	LECTURE/ WEEK
RPSCSP30	RPSCSP301: PRACTICAL OF Social Network Analysis & Simulation	2	4
	 Write a program to compute the following for a given a network: number of edges, number of nodes; degree of the node; node with the lowest degree; the adjacency list; matrix of the graph. Length of the shortest path edge list Perform the following tasks: View data collection forms and/or import one-mode/two-mode datasets; Basic Networks matrices transformations Compute the following node level measures: 		

	 Density; Degree; Reciprocity; Transitivity; Centralization; Clustering. equivalence create sociograms for the persons-by-persons network and the committee-by-committee network for a given relevant problem. Create a one-mode network and two-node network for the same. Bipartite Graph 		LC-9C
5.	node network for the same.	0	
	Hamming distance, Manhattan, Euclidean		
	Perform SVD analysis of a network.		
8.	Case studies		

COURSE CODE	COURSE NAME	CREDITS	LECTURE /WEEK
RPSCS302	RPSCS302: Internet of Things	4	4

Course Objective:

- Work on cloud-based Embedded Systems
- Explore various aspects of IOT
- Explore and understand different technologies associated with IOT

Expected Learning Outcome:

- Implement various IOT based project in Business and Healthcare
- Exploit the IOT Technologies for the benefit of humans and to solve problems today.

UNITS	COURSE CONTENTS	NO. OF LECTURE S
I	Introduction to internet of things: Introduction, Physical Design of IOT, IOT Enabling Technologies, IOT Levels and deployment templates, Domain-Specific IOT: Home IOT, IOT in Cities, IOT in Environment, IOT used for Energy, IOT in Health and lifestyle	15 L
. 10.30	IOT and M2M, Difference between IOT and M2M, SDN and NFV for IOT, Software-defined Networking, Function visualization, IOT Management with NETCONF-YANG: Need for IOT System management, Simple Network Management Protocol, Network Operator requirement, NETCONF, YANG, IOT System management with NETCONF, YANG	
	IOT Physical Server and Cloud offerings: Introduction to cloud storage model and Communication API, WAMP- AutoBahn for IOT, Google Cloud for IOT, Python web application framework Django, Designing restful web API, Amazon Webservices for IOT, Skynet IOT Messaging platform.	
	Case Studies: Introduction, Home Automation, Cities, Environments, Agriculture, Productive applications, health care, automobiles.	15 L

Text book:

- 1. Google IOT cloud
- 2. Internet of Things: A hands-on Approach by Arshdeep Bahga and Vijay Madisetti

COURSE CODE	COURSE NAME	CREDITS	LECTURE/ WEEK
RPSCSP302	RPSCSP302: PRACTICAL OF Internet of Things	2	4
1	Creating Pub/Sub Topic		
2	Creating registries and devices		
3	Creating cloud Storage bucket	60	
4	Creating a cloud dataflow pipeline		
5	streaming IOT data to cloud storage		
6	Streaming IOT data to Big Query	3	
7	Streaming IOT data to Cloud Prep		
8	Innovative challenge lab.		

COURSE CODE	COURSE NAME	CREDITS	LECTURE /WEEK
RPSCS303	RPSCS303: Big Data Technologies	4	4

Course Objective:

- Understanding various Big Data Technologies
- Working with R and Hadoop

Expected Learning Outcome:

Students completing this course will be able to:

• Develop applications in Hadoop and R for Big Data

UNITS	COURSE CONTENTS	NO. OF LECTURE S
SIMI	Installing R, Installing RStudio, Understanding the features of the R language, Installing Hadoop, Understanding Hadoop features, Learning the HDFS and MapReduce architecture, Understanding Hadoop subprojects, Writing Hadoop MapReduce Programs, Understanding the basics of MapReduce, Introducing Hadoop MapReduce, Understanding the Hadoop MapReduce fundamentals, Writing a Hadoop MapReduce example, Learning the different ways to write Hadoop MapReduce in R.	
	Integrating R and Hadoop: Introducing RHIPE, Understanding the architecture of RHIPE, Understanding RHIPE samples, Introducing RHadoop. Using Hadoop Streaming with R: Understanding the basics of Hadoop streaming, understanding how to run Hadoop streaming with R	

III	Learning Data Analytics with R and Hadoop: Understanding the data analytics project life cycle, Understanding data analytics problems, Exploring web pages categorization, Computing the frequency of stock market change, Predicting the sale price of the blue book for bulldozers – case study	
	Importing and Exporting Data from Various DBs: Learning about data files as the database, Understanding MySQL, Understanding Excel, Understanding MongoDB, Understanding SQLite, Understanding PostgreSQL, Understanding Hive, Understanding HBase.	

Text book:

- 1. Big Data Analytics with R and Hadoop -Set up an integrated infrastructure of R and Hadoop to turn your data analytics into Big Data analytics Vignesh Prajapati
- 2. Hadoop: The Definitive Guide by Tom White 5th Edith O'Reilly.

COURSE CODE	COURSE NAME	CREDITS	LECTURE/ WEEK
RPSCSP303	PRACTICAL OF RPCSP303 Bigdata Technologies	2	4
1	Implement Mapreduce for a case study		
2	Demonstrate an application which uses RHIPE.		
3	Demonstrate a streaming application using RHadoop		
	Demonstrate an application in R to implement predictive methods in data analytics.		
5	Demonstrate usage of HIVE in Hadoop		
6.	Demonstrate the usage of HBase		
7.	Extract data from various sources and load it in Hadoop		
8.	Demonstrate the data visualization using R		

COURSE CODE	COURSE NAME	CREDITS	LECTURE /WEEK
RPSCS401	RPSCS401 Simulation & Modeling	4	4

Learning Objective:

- Developing and understanding various simulation models
- Understand the application of simulation models to perform projections

Learning Outcome:

• Develop skills for encoding a given problem and understanding which statistical technique will help stimulate the same.

UNITS	COURSE CONTENTS	NO. OF LECTURE S
I	Introduction to Simulation: System and System environment, Components of system, Type of systems, Type of models, Steps in the simulation study, Advantages and Disadvantages of simulation. Simulation Examples: Simulation of Queueing systems, Other examples of simulation. Concepts of discrete event simulation, List processing, History of simulation software, Desirable software features, General-purpose simulation packages, Object-oriented simulation, Trends in simulation software Simulation Modeling: The need for Simulation, Time to simulate, Inside simulation software: Modeling the progress of Time, Modeling Variability	15 L
II	Statistical Models in Simulation: Useful statistical model, Discrete distribution, Continuous distribution, Poisson process, Empirical distribution. Queueing Models: Characteristics of Queueing systems, Queueing notations, Long run measures of performance of Queueing systems, Steady-state behaviour of infinite population Markovian models, Steady-state behaviour finite population model, Network of Queues. Random Number Generation: Properties of random numbers, Generation of pseudo-random numbers, Techniques for generating random numbers, Tests for random numbers. Inverse transform technique, Convolution method, Acceptance rejection techniques	15 L
III	Conceptual Modeling: Introduction to Conceptual modelling, Defining conceptual model, Requirements of the conceptual model, Communicating the conceptual model, Developing the Conceptual, A framework for conceptual modelling, methods of model simplification. Types of models, Analytical vs Simulation modelling, Application of simulation modelling, Agent-Based Modelling, Designing state-based behaviour	15 L
IV	Model Verification and Validation	15 L

Data Collection and Analysis: Introduction, Data requirements, Obtaining data, Representing unpredictable variability, Selecting statistical distributions. Obtaining Accurate Results: Introduction, The nature of simulation models and simulation output, Issues in obtaining accurate simulation results, example model, dealing with initialization bias: warm-up and initial conditions, Selecting the number of replications and run-length. Searching the Solution Space: Introduction, The nature of simulation experimentation, Analysis of results from a single scenario, Comparing alternatives, Search experimentation, and Sensitive analysis. Verification Validation and Confidence: Introduction, Defining Verification and Validation, The difficulties of verification and validation, Methods of verification and validation, Independent verification and validation.

- 1. Jerry Banks, John Carson, Barry Nelson, David Nicol, Discrete Event System
- 2. Simulation.Introduction to Social Network Methods: Robert A. Hanneman, Mark Riddle, University of California, 2005 [Published in digital form and available athttp://faculty.ucr.edu/~hanneman/nettext/index.html].
- 3. Simulation: The Practice of Model Development and Use by Stewart Robinson, John Wiley and Sons, Ltd, 2004.
- 4. The Big Book of Simulation Modeling: Multimethod Modeling by AndreiBorshchev, 2013

COURSE	COURSENAME		LECTUR
CODE		S	E/WEEK
RPSCSP401	RPSCSP401 PRACTICAL OF Simulation & modeling	2	4
1	Design and develop an agent-based model		
2.	Design and develop the System Dynamic model		
3.	Design and develop a discrete-event model that will simulate the process.		
4.	Design and develop time-slice simulation for a scenario like an airport model to design how passengers move within a small airport that hosts two airlines, each with their own gate. Passengers arrive at the airport, check-in, pass the security checkpoint and then go to the waiting area. After boarding starts, each airline's representatives check their passengers' tickets before they allow them to board.		
5.	 Verify and validate a model developed like a bank model or manufacturing model & Create a defense model to simulate aircraft behaviour. 		
6.	Statistical models		-
7.	queuing models		
8.	random number generation		
9.	Case studies		

M.Sc Par

Internal Exam - 40 Marks

1. 20 Marks -- MCQ Test:

Tests will be taken based on any of the unit/units as informed by the faculty in-charge.

2. 20 Marks -

Develop Mini project in a group(max four) and presentation of the same. / Online Course.

Note: Students have to acquire at least 40% marks in each paper individually. Callege

External Examination - 60 Marks Duration 21/2 Hrs

Theory Question Paper Pattern:

All Questions are Compulsory					
Questions	Based On	Marks			
Q1	Unit I	12			
Q2	Unit II	12			
Q3	Unit III	12			
Q4	Unit IV	12			
Q5	MIX	12			

- All questions are compulsory with internal choice within the questions.
- Each Question will have 3 sub-questions carrying 6 marks each, out of which student has to answer any 2.

Note: Students have to acquire at least 40% marks in each paper individually.

Internal Exam - 20 Marks

1. Innovative Practical -- 10 Marks

- It can be clubbed with a mini project as an additional application.
- 2. Give a separate application based on the theory paper.

Regularity -- 10 Marks

Timely submission of practical's on the google classroom.

- 2. Attendance should be 75%.
- 3. Submission of the e-journal on time.

Note: Students have to acquire at least 40% marks in each paper individually.

External Examination - 30 Marks Practical Question -

- 1 or 2 questions can be asked in the practical exam for each paper.
- External will be called for evaluating the same.

Note:

- 1. Students who have submitted their e-Journal and certified are only allowed to ar indiv appear for the exam.
 - 2. Students have to acquire at least 40% marks in each paper individually.

MSc Part II (Sem III & IV) EVALUATION SCHEME

THEORY

Internal Exam - 40 Marks

1. **20 Marks -- MCQ Test:**

Tests will be taken based on any of the unit/units as informed by the faculty in-charge.

1. **20 Marks –**

Assignments based on the syllabus or any other topic in demand based on the syllabus.

Note: Students have to acquire at least 40% marks in each paper individually.

External Examination - 60 Marks Duration 21/2 Hrs

Theory Question Paper Pattern:-

All Questions are Compulsory					
Questions	Based On	Marks			
Q1	Unit I	12			
Q2	Unit II	12			
Q3	Unit III	12			
Q4	Unit IV	12			
Q5	MIX	12			

- All questions shall be compulsory with internal choice within the questions.
- Each Question will have 3 sub-questions carrying 6 marks each, out of which student has to answer any 2.

Note: Students have to acquire at least 40% marks in each paper individually.

PRACTICAL

Internal Exam - 20 Marks

1. Innovative Practical -- 10 Marks

- 1. It can be clubbed with assignments.
- 2. Research paper review to be done for an application.

2. Regularity -- 10 Marks

- 1. Timely submission of practical's on the google classroom.
- 2. Attendance should be 75%.
- 3. Submission of the e-journal on time.

Note: Students have to acquire at least 40% marks in each paper individually.

External Examination - 30 Marks Practical Question -

- 1 or 2 questions can be asked in the practical exam for each paper.
- External will be called for evaluating the same.

Note:

- 1. Students who have submitted their e-Journal and certified are only allowed to appear for the exam.
- 2. Students have to acquire at least 40% marks in each paper individually.

PROJECT EVALUATION - 150 Marks

Internal evaluation - 60 Marks

- Abstract submission & literature Survey / sample data collection 10 Marks
- Technology Implementation 10 marks
- Mid-Term Presentation 20 Marks
- Project Documentation- 20 marks

External evaluation - 90 Marks

- Project Quality 20 Marks.
- Project Implementation 40 Marks.
- Presentation 30 Marks.

A student should submit a project implementation report with the following details:

- 1. **Title:** Title of the project.
- 2. **Implementation details:** A description of how the project has been implemented. It shall be of 2 to 4 pages.
- 3. **Experimental setup and results:** A detailed explanation of how experiments were conducted, what software used and the results obtained. Details like screenshots, tables and graphs can come here. It shall be of 6 to 10 pages.
- 4. **Analysis of the results:** A description of what the results mean and how they have been arrived at. Different performing measures or statistical tools used etc may be part of this. It shall be of 4 to 6 pages.
- 5. **Conclusion:** A conclusion of the project performed in terms of its outcome (Maybe half a page).
- 6. **Future enhancement:** A small description on what enhancement can be done when more time and resources are available (Maybe half a page).
- 7. **Program code:** The program code may be given as an appendix.

Note:

- 1. Students have to acquire at least 40% marks in project evaluation.
- 2. Internal evaluation will be done by the Project guide allotted.
- 3. Timely submission on google classroom as per requirement is must, regularity will be determined based on that.
- 4. Attendance should be 75%.

INTERNSHIP EVALUATION - 450 Marks

Internal evaluation - 180 Marks

Following are the guidelines for evaluation:

- 1. Job description: 20 Marks
- 2. Technical knowledge/skills: 40 Marks
- 3. Innovation & creativity: 40 Marks
- 4. Adherence to Schedule (weekly activity report): 20 Marks
- Soft Skills (Communication, Teamwork, Resource Management, Leadership qualities):
 40 Marks
- 6. Discipline & behaviour: 20 Marks

External evaluation - 270 Marks (50% by employer & 50% by external examiner)

On the basis of the detailed internship report submitted by the student duly signed by the employer and the internal faculty. A presentation is expected from the student for sharing his / her learning experience and work done at the internship.

Following are the guidelines for evaluation:

- 1. Internship Report: 30 Marks
- 2. Innovation and creativity: 50 Marks
- 3. Experience-based learning: 50 Marks
- 4. Viva: 20 Marks
- 5. Internship Genuineness: 20 Marks
- 6. Soft Skills: 30 Marks
- 7. Suitability & Clarity of material presented: 30 Marks
- 8. Quality of oral presentation: 40 Marks

Note: - Students need to find Internship by themselves. It's their responsibility.

Following things are expected to be completed by the student for the final evaluation.

- 1) The syllabus proposes an internship for about 600 hours to be done by a student.
- 2) It is expected that a student chooses an IT or IT-related industry and formally works as a full-time intern during the period.
- 3) The evaluation will be done based on the feedback given by the employers about the student.
- 4) The student should subject oneself with an internship evaluation with proper documentation of the attendance and the type of work he or she has done in the chosen organization.

Following are the guidelines laid for the same

- 1) Internship joining Letter with a proper job description.
- 2) Weekly Report in Excel format to be shown every week to Internal In-charge
 - Start date
 - End date
 - Task Assigned

- Task completed
- Outcome / Learning's
- 3) Internship completion Letter with proper hours & task completed.
- 4) Employer Feedback Form is prepared to assess based on the following:
 - Skills/ Knowledge
 - Self-Management
 - Dependability
 - Attitude
 - Relationships
- 5) Internship report:
 - Organization Overview
 - Description (Role, Activities, Technology Used, Live project link or screenshots)
 - SWOT Analysis
 - Introspection (knowledge acquired, Skills learned, challenging tasks performed)
 - Employers Feedback.
- 6) Proper certification by the person, to whom the student was reporting, with Organization's seal should be attached as part of the documentation.

PASSING CRITERIA 40%: - Students must acquire a minimum of 40% marks in each course individually (Theory, Practical, Project & Internship).