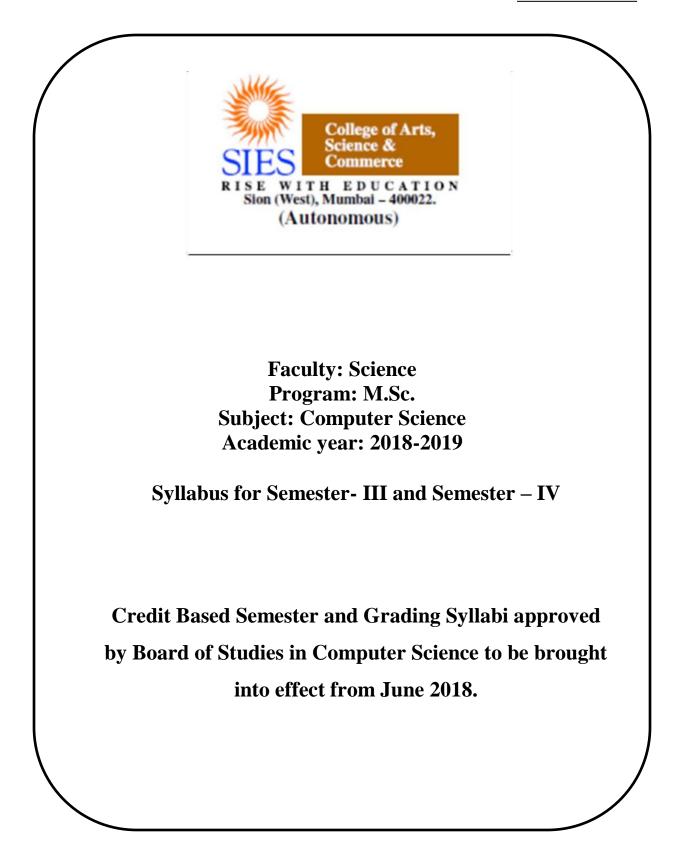


NAAC REACCREDITED - 'A' GRADE BEST COLLEGE AWARD - UNIVERSITY OF MUMBAI

Academic Council: --/--/--Item No. :



Preamble

This syllabus is an extension of the syllabus for semester - I and semester - II of MSc Computer Science, which came into existence in the academic year 2018-2019. As mentioned in the syllabus of semester I and II, the intended philosophy of the new syllabus is to meet following guidelines:

- Give strong foundation on core Computer Science subjects.
- Expose student to emerging trends in a gradual and incremental way.
- Prepare student community for the demands of ICT industry.
- Offer specialization on a chosen area.
- Create research temper among students in the whole process.

This syllabus for the semester - III and semester – IV has tried to continue the steps initiated in the semester- I and semester –II to meet the goals set. This proposes two core compulsory subjects in semester III. The student has to continue with the tracks they have taken in the semester II as elective subjects. The syllabus also includes project proposal as part of the practical course in elective subjects.

The semester – IV will have one compulsory subject. Student can choose one subject as specialization out of the two electives he or she has been pursuing since the semester – II. That means, there will be four specializations in the semester IV as mentioned below:

- Cloud Computing
- Cyber and Information Security
- Business Intelligence and Big Data Analytics
- Machine Learning

The syllabus also offers an internship and project implementation in the semester – IV, each of which has weights equivalent to a full course. By introducing different electives as tracks in semester –II, espousing more of that tracks in the semester –III and offering the opportunity to choose the specialization based on the tracks pursed in semester –IV 3

will give the student the added advantage of high level competency in the advanced and emerging areas of computer science. This will definitely equip the student with industry readiness as internship in an IT or IT-related organization gives a practical exposure to what is learned and what is practiced. The strong foundation given in the core courses in different semesters will give enough confidence to the learner to face and adapt to the changing trends and requirements of industry and academia.

As one can easily notice, the syllabus offers lots of emphasis on student driven learning and learning through experience. Research is embedded in the course structure. By introducing Researching Computing in semester – I, Case study in semester – II, Project Proposal in semester – III and Project Implementation in semester – IV (which together has a weightage equivalent to almost two theory courses), the syllabus prepares a strong army of budding computer science researchers. The syllabus designed on the firm believe that by focusing on student driven research on cutting edge and emerging trends with lots of practical experience will make the learning more interesting and stimulating. It is hoped that the student community and teacher colleagues will appreciate the thrust, direction and treatment given in the syllabus.

We wholeheartedly thank all experts who shared their valuable feedbacks and suggestions to improvise the contents, we have sincerely attempted to incorporate each of them. We further thank Chairperson and members of Board of Studies for their confidence in us. Special thanks to Department of Computer Science and colleagues from various colleges, who volunteered or have indirectly helped designing certain specialized courses and the syllabus.

Structure of the syllabus

This is the syllabus for the semester–III and semester–IV of MSc Computer Science program to be implemented from the year 2018-2019.

Semester-III

The syllabus offers four theory courses and two practical courses in semester-III. Of the four theory courses, two are compulsory courses. The remaining two are electives. Each elective course has two tracks (track A and track B for elective I and track C and track D for elective II). A student is expected to continue with the track they have chosen in semester-II.

The syllabus proposes four subjects in semester-III. Each subject has theory and practical components.

Semester-III: Theory courses

The four theory courses offered in semester-III are:

- (i) Ubiquitous Computing
- (ii) Social Network Analysis
- (iii) Elective I
 - (a) Track A: Cloud Computing II (Cloud Computing Technologies)
 - (b) Track B: Cyber and Information Security II (Cyber Forensics)
- (iv) Elective II
 - (a) Track C: Business Intelligence and Big Data Analytics II (Mining Massive Data sets)
 - (b) Track D: Machine Learning II (Advanced Machine Learning)

A student is expected to continue with the same tracks he or she has taken in semester-II for elective –I and elective –II. Each of these theory courses (compulsory as well as elective) is of four credits each and is expected to complete in 60 hours. The details are shown in the following table.

Course	Course	Lecture	Credits
Code	Nomenclature	In Hours	
SIPSCS31	Ubiquitous Computing	60	4
SIPSCS32	Social Network Analysis	60	4
010000004	Elective I - Track A: Cloud Computing –II		
SIPSCS331	(Cloud Computing Technologies)		
	Elective I - Track B: Cyber and	60	4
SIPSCS332	Information Security- II (Cyber Forensics)		
	Elective II - Track C: Business Intelligence		
SIPSCS341	and Big Data Analytics –II		
	(Mining Massive Data sets)	60	4
SIPSCS342	Elective II - Track D: Machine Learning -II	-	
	(Advanced Machine Learning)		
Total Credits for Theory courses in Semester III		16	

Semester III – Theory courses

Semester-III: Practical Laboratory Courses

The syllabus proposes two laboratory courses of 4 credits each. The laboratory experiments from the first two theory courses (SIPSCS31 and SIPSCS32) are combined together and are proposed as the first practical course (SIPSCSP5). Similarly, the laboratory experiments from the elective courses are combined together and taken as the second practical course (SIPSCSP6). The following table summarizes the details of the practical courses in the semester –III.

Course code	Course Title	No of hours	Credits
SIPSCSP5	Ubiquitous Computing and Social Network	60+60=	04
	Analysis	120	
SIPSCSP6	Elective I and Elective II	60+60=	04
		120	
Total (Predits for Practical Laboratory courses in Semester	-111	08

Semester-III: Practical Laboratory Courses

Project Proposal: The syllabus introduces a project proposal in the semester-III under lab course SIPSCSP6. As per this, a student is expected to select a topic for project based on the specialization he or she is planning to take in the semester-IV. Needless to say, the project proposal will be based on a topic related to the elective the student has been pursuing in semester –II and semester-III and intends to continue in semester- IV as specialization.

The proposal will contain introduction, related works, objectives and methodology. The implementation, experimental results and analysis will be part of the Project implementation in the semester-IV.

Semester –IV

The syllabus proposes two subjects in semester-IV, each with theory and practical components. In addition, there will be internship with industry and a project implementation. The important feature of the semester-IV is the specialization a student can choose. A student can choose a specialization based on the electives one has been pursuing since semester–II. Since there are two electives in semester-III, a student can drop one and choose the other as the specialization in semester–IV.

Semester–IV: Theory courses

The two theory courses offered in semester-IV are:

- (i) Simulation and Modeling
- (ii) Specialization
 - (a) Track A: Cloud Computing III (Building Clouds and Services)
 - (b) Track B: Cyber and Information Security–III (Cryptography and Crypt Analysis)
 - (c) Track C: Business Intelligence and Big Data Analytics III (Intelligent Data Analysis)
 - (d) Track D: Machine Learning III (Computational Intelligence)

Each of these courses (core as well as the specialization) is expected to complete in 60 hours. The details are given in the following table.

Course Code	Course	Lecture	Credits
	Nomenclature	In Hours	
SIPSCS41	Simulation and Modeling	60	4
SIPSCS421	Specialization - Track A: Cloud Computing –III		
317303421	(Building Clouds and Services)		
	Specialization - Track B: Cyber and Information		
SIPSCS422	Security- II (Cryptography and Crypt Analysis)	60	4
SIPSCS423	Specialization - Track C: Business Intelligence and		
SIF 303423	Big Data Analytics –III (Intelligent Data Analysis)		
SIPSCS424	Specialization - Track D: Machine Learning –III		
3IF 303424	(Computational Intelligence)		
	Total Credits for Theory courses in Semester-IV		08

Semester-IV: Theory courses

Semester–IV: Practical Laboratory courses

The syllabus proposes one laboratory course of 4 credits. The laboratory experiments from the two theory courses are combined together and are proposed as the first practical course (PSCSP7).

Course code	Course Title	No of hours	Credits
SIPSCSP7	Simulation & Modeling and Specialization	60+60=	04
		120	

Semester-IV: Practical course

Semester-IV: Internship with industry

The syllabus proposes an internship for about 8 weeks to 12 weeks to be done by a student. It is expected that a student chooses an IT or IT-related industry and formally works as a full time intern during the period. 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. Proper certification (as per the guidelines given in Appendix 1 and 2) by the person, to whom the student was reporting, with Organization's seal should be attached as part of the documentation.

Semester-IV: Internship

Course code	Course Title	No of hours	Credits
SIPSCSP8	Internship with industry	300	06

Semester-IV: Project Implementation

The syllabus proposes project implementation as part of the semester–IV. The project implementation is continuation of the project proposal the students has submitted and evaluated in semester-III. The student is expected to continue with the proposal made and examined in the semester-III and implement the same in the semester–IV. In addition, experimental set up, analysis of results, comparison with results of related works, conclusion and future prospects will be part of the project implementation. A student is expected to make a project implementation report and appear for a project viva. He or she needs to spend around 200 hours for the project implementation, which fetches 6 credits. The details are given below:

Course code	Course Title	No of hours	Credits
SIPSCSP9	Project Implementation	200	06

Semester-IV: Project Implementation

Detailed syllabus of semester- III

Course Code	Course Title	Credits
SIPSCS31	Ubiquitous Computing	04
Unit I: Basics	of Ubiquitous Computing	
	biquitous Computing Applications, Holistic Framework for Ubi	Com: Smart
•	the Key Ubiquitous Computing Properties, Ubiquito	
Environment In	teraction, Architectural Design for UbiCom Systems: Smart	DEI Model,
	and Services, Service Architecture Models, Service Provision	
Unit II: Smart I	Mobiles, Cards and Device Networks	
	Devices, Users, Resources and Code, Operating Systems	s for
	ers and Communicator Devices, Smart Card Devices, Device	
•	uter Interaction (HCI): Explicit HCI, Implicit HCI, User Inte	
•	Devices, Hidden UI Via Basic Smart Devices, Hidden UI Via V	
and Implanted	Devices, Human Centered Design (HCD).	
Unit III: Smart	Environments	
Tagging, Sens	sing and Controlling, Tagging the Physical World, Se	ensors and
Sensor Networ	ks, Micro Actuation and Sensing: MEMS, Embedded Sy	stems and
Real Time Sys	tems, Control Systems.	
Unit IV: Ubiqu	itous Communication	
Audio Netwo	rks, Data Networks, Wireless Data Networks, Univ	versal and
Transparent A	udio, Video and Alphanumeric Data Network Access,	Ubiquitous
Networks, Networ	work Design Issues.	
Text book:		
Ubiquitou	is Computing Smart Devices, Environments and Interactio	ns, Stefan
Poslad, V	Viley, 2009.	
References:		
Ubiquitou	is Computing Fundamentals. John Krumm, Chapman & Hall/C	RC 2009.
Ambient	intelligence, wireless networking and ubiquitous computing],
	s, A., & Pedrycz, W. ArtechHouse, Boston, 2006.	-

• • http://www.eecs.qmul.ac.uk/~stefan/ubicom.

Course Code	Course Title	Credits
SIPSCS32	Social Network Analysis	04
	ction to social network analysis (SNA) networks and relations- analyzing relationships to understand	people and
groups, binary	and valued relationships, symmetric and asymmetric re	ationships,
multimode rel	lationships, Using graph theory for social networks	analysis-
adjacency matr	rices,edge-lists, adjacency lists, graph traversals and	distances,
depth-first trave	ersal, breadth-first traversal paths and walks, Dijkstra's	algorithm,
graph distance	e and graph diameter, social networks vs. link analysis,	ego-centric
and socio-cent	ric density.	
external and visualizing ego and betweenned centers, notic Analyzing network clans, K-plexed	networks- density, reachability, connectivity, reciproc group-internal ties in networks, ego networks, extra o networks, structural holes, Centrality- degree of centrality ess centrality, local and global centrality, centralization on of importance within network, Google pagerank work structure- bottom-up approaches using cliques, N- es, K-cores, F-groups and top-down approaches using c points, lambda sets and bridges, and factions.	acting and , closeness and graph algorithm, cliques, N-
Unit III: Measu	res of similarity and structural equivalence in SNA	
Approaches to	o network positions and social roles- defining equiv	valence or
similarity, struc	ctural equivalence, automorphic equivalence, finding equiva	alence sets,
brute force and	d Tabu search, regular equivalence, equivalence of	distances:
Maxsim, regu	lar equivalence, Measuring similarity/dissimilarity- valued	relations,
Pearson corre	elations covariance and cross-products, Understanding	clustering-
agglomerative	and divisive clusters, Euclidean, Manhattan, and squared	distances,
binary relations	s, matches: exact, Jaccard, Hamming,	
Unit IV: Two-m	node networks for SNA	
Understanding	mode networks- Bi-partite data structures, visualizing	two-mode
data, quantitati	ve analysis using two-mode Singular value decomposi	tion (SVD)

two-mode factor analysis, two-mode correspondence analysis, qualitative analysis using two-mode core-periphery analysis, two-mode factions analysis, affiliation and attribute networks.

Text book:

- Introduction to Social Network Methods: Robert A. Hanneman, Mark Riddle, University of California, 2005 [Published in digital form and available at http://faculty.ucr.edu/~hanneman/nettext/index.html].
- Social Network Analysis for Startups- Finding connections on the social web: Maksim Tsvetovat, Alexander Kouznetsov, O'Reilly Media, 2011.
- Social Network Analysis- 3rd edition, John Scott, SAGE Publications, 2012.

Reference book:

- Exploratory Social Network Analysis with Pajek, Second edition: Wouter de Nooy, Andrej Mrvar, Vladimir Batagelj, Cambridge University Press, 2011.
- Analyzing Social Networks, Stephen P Borgatti, Martin G. Everett, Jeffrey C. Johnson, SAGE Publications, 2013.
- Statistical Analysis of Network Data with R: Eric D. Kolaczyk, Gábor Csárdi, Springer, 2014.
- Network Analysis: Methodological Foundations, (Editors) Ulrik Brandes, Thomas Erlebach. Springer, 2005.
- Models and Methods in Social Network Analysis: (Editors) Peter J.
 Carrington, John Scott, Stanley Wasserman, Cambridge University Press, 2005.

Course Code	Course Title	Credits
SIPSCS331	Elective I- Track A: Cloud Computing -II	04
	(Cloud Computing Technologies)	
	and Distributed Computing arallel computing, elements of distributed computing, Tech	nnologies
for distributed c	computing: RPC, Distributed object frameworks, Service orient	ted
computing Virtu	alization – Characteristics, taxonomy, virtualization and cloud	computing.
Cloud Computi	ting Platforms ng definition and characteristics, Enterprise Computing, The ir	
•	computing services: SaaS, PaaS, IaaS, Enterprise architectu	ire, Types
of clouds.		
Unit III: Cloud Cloud computi	Technologies ng platforms, Web services, AJAX, mashups, multi-tenant	software,
Concurrent co	mputing: Thread programming, High-throughput comput	ing: Task
programming, [Data intensive computing: Map-Reduce programming.	
	are Architecture ns, Enterprise software: ERP, SCM, CRM	
Custom enterp	ise applications and Dev 2.0, Cloud applications.	
Text book:		
	e Cloud Computing Technology, Architecture, Applications, G ambridge University Press, 2010	Bautam
	g In Cloud Computing, Rajkumar Buyya, Christian Vecchiola / Selvi S, Tata Mcgraw-Hill Education, 2013	And
Cloud Co 2009	mputing: A Practical Approach, Anthony T Velte, Tata Mcgrav	v Hill,
References:		
(SaaS, Pa	ng the Cloud: Design Decisions for Cloud Computing Service aaS, and IaaS), Michael J. Kavis, Wiley CIO, 2014 mputing: SaaS, PaaS, IaaS, Virtualization, Business Models,	

 Cloud Computing: SaaS, PaaS, IaaS, Virtualization, Business Mo Security and More, Kris Jamsa, Jones & Bartlett Learning, 2013

Course Code	Course Title	Credits
SIPSCS332	Elective I- Track B: Cyber and Information Security- II	04
	(Cyber Forensics)	
Unit I: Compu	uter Forensic Fundamentals: Introduction to Computer	r Forensics
and objective,	the Computer Forensics Specialist, Use of Computer	Forensic in
Law Enforceme	nt, Users of Computer Forensic Evidence, Case	e Studies,
Information Sec	curity Investigations. Types of Computer Forensics	Technology:
Types of Milita	ary Computer Forensic Technology, Types of Law E	Inforcement
Computer For	ensic Technology, Types of Business Computer Forensic	Technology,
Specialized For	ensics Techniques, Hidden Data, Spyware and	Adware,
Encryption	Methods and Vulnerabilities, Protecting Data fr	om Being
Compromised,	Internet Tracing Methods, Security and Wireless Te	chnologies.
Types of Com	nputer Forensics Systems: Study different Security Syste	m: Internet,
Intrusion Detect	tion, Firewall, Storage Area, Network Disaster Recovery,	Public Key
Infrastructure,	Wireless Network, Satellite Encryption, Instant Messagir	ng (IM), Net
Privacy, Identity	Management, Biometric, Identity Theft.	
Unit II: Data I	Recovery: Data Recovery and Backup, Role of Data Recovery	very, Hiding
and Recovering	Hidden Data. Evidence Collection: Need to Collect the Evide	ence, Types
of Evidences,	The Rules of Evidence, Collection Steps. Compu	iter Image
Verification and	Authentication: Special Needs of Evidence Aut	hentication.
Identification of	of Data: Timekeeping, Forensic Identification and A	nalysis of
Technical Surve	eillance Devices, Reconstructing Past Events: How to Becon	me a Digital
Detective, Usea	ble File Formats, Unusable File Formats, Converting Files.	
Unit III: Netwo	ork Forensics: Sources of Network Based Evidence, P	rinciples of
Internetworking,	Internet Protocol Suite. Evidence Acquisition: Physical I	nterception,
Traffic Acquisiti	ion Software, Active Acquisition. Traffic Analysis: Protoco	ol Analysis,
Packet Analysis	, Flow Analysis, Higher-Layer Traffic analysis. Statistical Flo	w Analysis:
Sensors, Flow F	Record Export Protocols, Collection and Aggregation, Analysi	s. Wireless:
	2 Protocol Series, Wireless Access Point, Wireless Traffic C	•
Analysis, Comm	ion Attacks, Locating Wireless Devices. Network Intrusion De	etection and 15

Analysis: NIDS/NIPS Functionality, Modes of Detection, Types of NIDS/NIPS, NIDS/NIPS Evidence Acquisition.

Unit IV: Network Devices and Mobile Phone Forensics: Sources of Logs, Network Architecture, Collecting and Analyzing Evidence, switches, routers, firewalls, interfaces Web Proxies: Need to Investigate Web Proxies, Functionality, Evidence, Squid, Web Proxy Analysis, Encrypted Web Traffic. Mobile Phone Forensics: Crime and Mobile Phones, Voice, SMS and Identification of Data Interception in GSM, Mobile Phone Tricks, SMS Security, Mobile Forensic.

Text book:

- Computer Forensics Computer Crime Scene Investigation, John R. Vacca, Second Edition, 2005.
- Network Forensics, Sherri Davidoff, Jonathan HAM, Prentice Hall, 2012.
- Mobile Phone Security and Forensic: A Practical Approach, Second Edition, Iosif Androulidkis, Springer, 2012.

- Digital forensics: Digital evidence in criminal investigation", Angus M.Marshall, John – Wiley and Sons, 2008.
- Computer Forensics with FTK, Fernando Carbone, PACKT Publishing, 2014.
- Practical Mobile Forensics, Satish Bommisetty, Rohit Tamma, Heather Mahalik, PACKT Publishing, 2014.

SIPSCS341 Elective I- Track C: Business Intelligence and Big Data Analytics –II (Mining Massive Data sets) 04 Unit I: Introduction To Big Data Big data: Introduction to Big data Platform, Traits of big data, Challenges of conventional systems, Web data, Analytic processes and tools, Analysis vs Reporting, Modern data analytic tools, Statistical concepts: Sampling distributions, Re-sampling, Statistical Inference, Prediction error. Data Analysis: Regression modeling, Analysis of time Series: Linear systems analysis, Nonlinear dynamics, Rule induction, Neural networks: Learning and Generalization, Competitive Learning, Principal Component Analysis and Neural Networks, Fuzzy Logic: Extracting Fuzzy Models from Data, Fuzzy Decision Trees, Stochastic Search Methods. Unit II: MAP REDUCE Introduction to Map Reduce: The map tasks, Grouping by key, The reduce tasks, Combiners, Details of MapReduce Execution, Coping with node failures. Algorithms Using MapReduce: Matrix-Vector Multiplication, Computing Selections and Projections, Union, Intersection, and Difference, Natural Join. Extensions to MapReduce: Workflow Systems, Recursive extensions to MapReduce, Common map reduce algorithms. Unit II: SHINGLING OF DOCUMENTS Finding Similar Items, Applications of Near-Neighbor Search, Jaccard similarity of sets, Similarity of documents, Collaborative filtering as a similar-sets problem, Documents, k- Shingles, Choosing the Shingle Size, Hashing Shingles, Shingles built from Words. Similarity-Preserving Summaries of Sets, Locality-Sensitive hashing for documents. The Theory of Locality-Sensitive functions. Methods for high degrees of similarity. Unit IV: MINING DATA STREAMS Introduction to streams concepts – Stream data model and architecture, Stream computing, Sampling data in a stream, Filtering streams, Counting distinct elements	Course Code	Course Title	Credits
Analytics –II (Mining Massive Data sets) Unit I: Introduction To Big Data Big data: Introduction to Big data Platform, Traits of big data, Challenges of conventional systems, Web data, Analytic processes and tools, Analysis vs Reporting, Modern data analytic tools, Statistical concepts: Sampling distributions, Re-sampling, Statistical Inference, Prediction error. Data Analysis: Regression modeling, Analysis of time Series: Linear systems analysis, Nonlinear dynamics, Rule induction, Neural networks: Learning and Generalization, Competitive Learning, Principal Component Analysis and Neural Networks, Fuzzy Logic: Extracting Fuzzy Models from Data, Fuzzy Decision Trees, Stochastic Search Methods. Unit II: MAP REDUCE Introduction to Map Reduce: The map tasks, Grouping by key, The reduce tasks, Combiners, Details of MapReduce Execution, Coping with node failures. Algorithms Using MapReduce: Matrix-Vector Multiplication, Computing Selections and Projections, Union, Intersection, and Difference, Natural Join. Extensions to MapReduce: Workflow Systems, Recursive extensions to MapReduce, Common map reduce algorithms. Unit II: SHINGLING OF DOCUMENTS Finding Similar Items, Applications of Near-Neighbor Search, Jaccard similarity of sets, Similarity of documents, Collaborative filtering as a similar-sets problem, Documents, k-Shingles, Choosing the Shingle Size, Hashing Shingles, Shingles built from Words. Similarity. Unit II: SHING DATA STREAMS Introduction to streams 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	SIPSCS341	Elective I- Track C: Business Intelligence and Big Data	04
 Big data: Introduction to Big data Platform, Traits of big data, Challenges of conventional systems, Web data, Analytic processes and tools, Analysis vs Reporting, Modern data analytic tools, Statistical concepts: Sampling distributions, Re-sampling, Statistical Inference, Prediction error. Data Analysis: Regression modeling, Analysis of time Series: Linear systems analysis, Nonlinear dynamics, Rule induction, Neural networks: Learning and Generalization, Competitive Learning, Principal Component Analysis and Neural Networks, Fuzzy Logic: Extracting Fuzzy Models from Data, Fuzzy Decision Trees, Stochastic Search Methods. Unit II: MAP REDUCE Introduction to Map Reduce: The map tasks, Grouping by key, The reduce tasks, Combiners, Details of MapReduce Execution, Computing Selections and Projections, Union, Intersection, and Difference, Natural Join. Extensions to MapReduce: Workflow Systems, Recursive extensions to MapReduce, Common map reduce algorithms. Unit II: SHINGLING OF DOCUMENTS Finding Similar Items, Applications of Near-Neighbor Search, Jaccard similarity of sets, Similarity of documents, Collaborative filtering as a similar-sets problem, Documents, k-Shingles, Choosing the Shingle Size, Hashing Shingles, Shingles built from Words. Similarity-Preserving Summaries of Sets, Locality-Sensitive hashing for documents. The Theory of Locality-Sensitive functions. Methods for high degrees of similarity. Unit IV: MINING DATA STREAMS Introduction to streams 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 			
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computing, Sampling data in a stream, Filtering streams, Counting distinct elements in a stream, Estimating moments, Counting oneness in a Window, Decaying window, Real	Finding Similar Similarity of doo Shingles, Cho Words. Simila documents. Th similarity. Unit IV: MINING	Items, Applications of Near-Neighbor Search, Jaccard similar cuments, Collaborative filtering as a similar-sets problem, Do osing the Shingle Size, Hashing Shingles, Shingles rity-Preserving Summaries of Sets, Locality-Sensitive I e Theory of Locality-Sensitive functions. Methods for high G DATA STREAMS	bouments, k- built from hashing for degrees of
	computing, San stream, Estima	npling data in a stream, Filtering streams, Counting distinct el ting moments, Counting oneness in a Window, Decaying w	lements in a

Text book:

- Mining of Massive Datasets, Anand Rajaraman and Jeffrey David Ullman, Cambridge University Press, 2012.
- Big Data, Big Analytics: Emerging Business Intelligence and Analytic Trends for Today's Businesses, Michael Minelli, Wiley, 2013

- Big Data for Dummies, J. Hurwitz, et al., Wiley, 2013
- Understanding Big Data Analytics for Enterprise Class Hadoop and Streaming Data, Paul C. Zikopoulos, Chris Eaton, Dirk deRoos, Thomas Deutsch, George Lapis, McGraw-Hill, 2012.
- Big data: The next frontier for innovation, competition, and productivity, James Manyika ,Michael Chui, Brad Brown, Jacques Bughin, Richard Dobbs, Charles Roxburgh, Angela Hung Byers, McKinsey Global Institute May 2011.
- Big Data Glossary, Pete Warden, O'Reilly, 2011.
- Big Data Analytics: From Strategic Planning to Enterprise Integration with Tools, Techniques, NoSQL, and Graph, David Loshin, Morgan Kaufmann Publishers, 2013

Course Code	Course Title	Credits
SIPSCS342	Elective I- Track D: Machine Intelligence - II	
	04 (Advanced Machine Learning Techniques)	
Unit I: Probabi A brief review	lity of probability theory, Some common discrete distributions	, Some
common conti	nuous distributions, Joint probability distributions, Transfor	mations of
random var	iables, Monte Carlo approximation, Information	theory.
Directed grap	hical models (Bayes nets): Introduction, Examples,	Inference,
Learning, Conc	litional independence properties of DGMs. Mixture model	s and EM
algorithm: Late	nt variable models, Mixture models, Parameter estimation	for mixture
models, The EN	A algorithm.	

Unit II: Kernels Introduction, kernel function, Using Kernel inside GLMs, kernel trick, Support vector machines, Comparison of discriminative kernel methods. Markov and hidden Markov models: Markov models, Hidden Markov Models (HMM), Inference in HMMs, Learning for HMMs. Undirected graphical models (Markov random fields): Conditional independence properties of UGMs, Parameterization of MRFs,Examples of MRFs, Learning, Conditional random fields (CRFs), applications of

CRFs.

Unit III: Monte Carlo inference

Introduction, Sampling from standard distributions, Rejection sampling, Importance sampling, Particle filtering, Applications: visual object tracking, time series forecasting, Rao-Blackwellised Particle Filtering (RBPF). Markov chain Monte Carlo (MCMC) inference: Gibbs sampling, Metropolis Hastings algorithm, Speed and accuracy of MCMC.

Unit IV: Graphical model structure learning

Structure learning for knowledge discovery, Learning tree structures, Learning DAG structure with latent variables, Learning causal DAGs, Learning undirected Gaussian graphical models, Learning undirected discrete graphical models. Deep learning: Deep generative models, Deep neural networks, Applications of deep networks.

Text book:

 Machine Learning: A Probabilistic Perspective: Kevin P Murphy, The MIT Press Cambridge (2012).

- Introducing Monte Carlo Methods with R, Christian P. Robert, George Casella, Springer, 2010
- Introduction to Machine Learning (Third Edition): Ethem Alpaydın, The MIT Press (2015).
- Pattern Recognition and Machine Learning: Christopher M. Bishop, Springer (2006)

• Bayesian Reasoning and Machine Learning: David Barber, Cambridge University Press (2012).

• Statistical And Machine Learning Approaches For Network Analysis, Edited By Practical Graph Mining with R: Edited by Nagiza-F-Samatova et al, CRC Press

(2014)

• https://class.coursera.org/pgm/lecture/preview

Cou	rse Code	Course Title	Credits
SIPSCSP31		Ubiquitous Computing	02
No		List of Practical Experiments	
1	Design a	and develop location based messaging app	
2	Design a	and develop chat messaging app which is a location-based	
3	Design a	and develop app demonstrating Simple Downstream Messagir	ng
4	Design and develop app demonstrating Send Upstream Messages		
5	Design and develop app for Device Group Messaging		
6	Implementing GCM Network Manager		
7	Demonstrate use of OpenGTS (Open Source GPS Tracking System)		
8	Context-Aware system		
	Context-awareness is a key concept in ubiquitous computing. The Java Context- Awareness Framework (JCAF) is a Java-based context-awareness infrastructure and programming API for creating context-aware applications		
9	Develop application demonstrating Human Computer Interaction		
10	Write a Java Card applet		

List of practical Experiments for Semester –III

Course Code Course Title			Credits	
SIPS	SCSP32	Social Network Analysis	02	
Sr				
No		List of Practical Experiments		
1	Write a program to compute the following for a given a network: (i) num		nber of	
	edges, (ii) number of nodes; (iii) degree of node; (iv) node with lowest degree;		degree; (v)	
	the adjac	ency list; (vi) matrix of the graph.		
2	Perform f	ollowing tasks: (i) View data collection forms and/or import or	າe-	
	mode/two	o-mode datasets; (ii) Basic Networks matrices transformation	S	
3	Compute	the following node level measures: (i) Density; (ii) Degree;		
	(iii) Recip	procity; (iv) Transitivity; (v) Centralization; (vi) Clustering.		
4	For a give	en network find the following: (i) Length of the shortest path fr	om a given	
	node to a	nother node; (ii) the density of the graph; (iii) Draw egocentri	c network of	
	node G w	vith chosen configuration parameters.		
5	Write a p	rogram to distinguish between a network as a matrix, a netwo	ork as an	
	edge list, and a network as a sociogram (or "network graph") using 3 distinct			
	networks	representatives of each.		
6	Write a p	rogram to exhibit structural equivalence, automatic equivaler	ce, and	
	regular e	quivalence from a network.		
7	Create s	ociograms for the persons-by-persons network and the comm	nittee-by-	
	committe	e network for a given relevant problem. Create one-mode ne	twork and	
	two-node	e network for the same.		
8	Perform	SVD analysis of a network.		
9	Identify ti	es within the network using two-mode core periphery analysi	S.	
10	Find "factions" in the network using two-mode faction analysis.			

Note:

One may use programming languages like R, Python, Pajek etc and open software/ tools like (i) EGONet; (ii) Ora; (iii) Netlogo; (iv) Pajek; and (v) NetDraw; to do the practical experiments.

Cou	rse Code	Course Title	Credits
SIPS	SCSP331	Practical Course on Elective I-Track A:Cloud	02
	1	Computing-II (Cloud Computing Technologies)	
Sr			
No		List of Practical Experiments	
1	Execute	& check the performance of existing algorithms using CloudS	sim.
2	Install a C	Cloud Analyst and Integrate with Eclipse/Netbeans. Monitor th	ne
	performa	nce of an Existing Algorithms.	
3	Build an application on private cloud.		
4	Demonst	rate any Cloud Monitoring tool.	
5	Evaluate a Private IAAS Cloud using TryStack.		
6	Implement FOSS-Cloud Functionality - VDI (Virtual Desktop Infrastructure)		cture)
7	Implemer	nt FOSS-Cloud Functionality VSI (Virtual Server Infrastructure	e)
	Infrastruc	ture as a Service (IaaS)	
8	Impleme	nt FOSS-Cloud Functionality - VSI Platform as a Service (Pa	aS)
9	Implement FOSS-Cloud Functionality - VSI Software as a Service (SaaS)		aS)
10	Explore F	FOSS-Cloud Functionality- Storage Cloud	

Cou	rse Code	Course Title	Credits
SIPSCSP332		Practical Course on Elective I-Track B: Cyber and	02
		Information Security- II (Cyber Forensics)	
Sr			
No		List of Practical Experiments	
1	Write a p	rogram to take backup of mysql database	
2	Write a p	rogram to restore mysql database	
3	Use Drive	elmage XML to image a hard drive	
4	Write a p	rogram to create a log file	
5	Write a p	rogram to find a file in a directory	
6	Write a p	rogram to find a word in a file	
7	Create for	prensic images of digital devices from volatile data such	as
	memory u	using Imager for: (i) Computer System; (ii) Server; (iii) Mobile	e Device
8		nd extract relevant information from Windows Registry for i using Registry View, perform data analysis and boo	Ū
	•	with respect to: (i) Computer System; (ii) Computer Network	
		v) Wireless Network	
9	Generate	a report based on the analysis done using Registry View for	different
	case scenario of the following: (i) Computer System; (ii) Computer Ne		er Network;
	(iii) Mobile	e Device; (iv) Wireless Network	
10	Create a	new investigation case using Forensic Tool: (i) Comput	er System;
	(ii) Comp	uter Network; (iii) Mobile Device ;(iv) Wireless Network.	

Cou	rse Code	Course Title	Credits
SIPSCSP341		Practical Course on Elective II-Track C: Business	02
		Intelligence and Big Data Analytics - II	
		(Mining Massive Data sets -I)	
No		List of Practical Experiments	
1	Generate	e regression model and interpret the result for a given data se	et.
2	Generate	e forecasting model and interpret the result for a given data se	et.
3	Write a	map-reduce program to count the number of occurrence	es of each
	alphabeti	c character in the given dataset. The count for each le	etter should
	be case-i	nsensitive (i.e., include both upper-case and lower-case	versions of
	the letter;	; Ignore non-alphabetic characters).	
4	Write a m	nap-reduce program to count the number of occurrences of e	each word in
	the give	en dataset. (A word is defined as any string of	alphabetic
	character	rs appearing between non-alphabetic characters like natu	ure's is two
	words.	The count should be case-insensitive. If a word occu	urs multiple
	times in	a line, all should be counted)	
5	Write a	map-reduce program to determine the average ratings	of movies.
	The inpu	t consists of a series of lines, each containing a movie ne	umber, user
	number,	rating and a timestamp.	
6	Write a	map-reduce program: (i) to find matrix-vector multiplica	ation; (ii) to
	compute	selections and projections; (iii) to find union,	intersection,
	difference	e, natural Join for a given dataset.	
7	Write a p	program to construct different types of k-shingles for given do	cument.
8	Write a	program for measuring similarity among documents an	d detecting
	passages	s which have been reused.	
9	Write a p	rogram to compute the n- moment for a given stream where	n is given.
10	Write a p	program to demonstrate the Alon-Matias-Szegedy Algorithm	for second
	moments		
Note	: The exp	periments may be done using software/tools like Hadoop	/ WEKA /
R / .	Java etc.		

Cou	rse Code	Course Title	Credits
SIPS	SCSP342	P342 Practical Course on Elective II- Track D: Machine	
		Intelligence - II (Advanced Machine Learning	
		Techniques)	
Sr			•
No		List of Practical Experiments	
1	Find p	robability density function or probability mas	s function,
		e distribution function and joint distribution function ies and quantiles for standard statistical distributions.	to calculate
2	Create a Directed Acyclic Graph (DAG) using (i) set of formulae (ii) set of vector and (iii) set of matrices. Find parents and children of nodes. Read conditional independence from DAG. Add and remove edges from graph.		ead
	-	ate a Bayesian network for a given narrative. Set findin One may use narratives like 'chest clinic narrative' and packa se].	•
4	Implemer	nt EM algorithm.	
5		ng kernel to find the similarity of two amino acid seque is defined as the number of a substring in common.	nce where
6	-	rate SVM as a binary classifier.	
7	Create a	random graph and find its page rank.	
8	Apply ran	dom walk technique to a multivariate time series.	
9	Implemer	nt two stage Gibbs Sampler.	
10	Implemer	nt Metropolis Hastings algorithm.	

Detailed syllabus of semester – IV

Course Code	Course Title	Credits
SIPSCS41	Simulation and Modeling	04
Unit I: Introduc	tion Simulation, Need of Simulation, Time to simulate, Inside	simulation
software: Mod	leling the progress of Time, Modeling Variability,	Conceptual
Modeling: Intr	oduction to Conceptual modeling, Defining conceptu	ual model,
Requirements of	of the conceptual model, Communicating the concept	ual model,
Developing the	e Conceptual Model: Introduction, A framework for	conceptual
U ,	hods of model simplification.	

Unit II: Model Verification and Validation

Data Collection and Analysis: Introduction, Data requirements, Obtaining data, Representing unpredictable Selecting variability, 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 The difficulties of verification validation, and Validation, and Methods of verification and validation, Independent verification and validation.

Unit III: Modeling and simulation modeling

Types of models, Analytical vs Simulation modeling, Application of simulation modeling, Level of abstraction, Simulation Modeling. Methods, System Dynamics, Discrete Event Modeling, Agent Based modeling: Introduction to Agent, Agent-based modeling, Time in agent based models, Space in agent based models, Discrete space, Continuous space movement in continuous space, Communication between agents, Dynamic creation and destruction of agents, Statics on agent population, Condition triggered events and transition in agents. Building agents based models: The problem statement, Phases of modeling, Assumptions, 3 D animation. Dynamics Systems: Stock and flow diagrams, examples of stock and flow diagrams. Multi-method modeling: Architecture, Technical aspects of combining modeling methods, Examples.

Unit IV: Design and behavior of models

Designing state-based behavior: Statecharts, State transitions, Viewing and debugging Statecharts at runtime, Statecharts for dynamic objects. Discrete events and Event model object: Discrete event, Event-the simplest low level model object, Dynamic events, and Exchanging data with external world. Presentation and animation: Working with shapes, groups and colors, Designing interactive models: using controls, Dynamic properties of controls, 3D Animation. Randomness in Models: Probability distributions, sources of randomness in the model, randomness in system dynamics model, random number generators, Model time, date and calendar: Virtual and real time: The model time, date and calendar, Virtual and real-time execution modes.

Text book:

- Simulation: The Practice of Model Development and Use by Stewart Robinson, John Wiley and Sons, Ltd, 2004.
- The Big Book of Simulation Modeling: Multi Method Modeling by Andrei Borshchev, 2013

- Agent Based Modeling and Simulation, Taylor S, 2014.
- Simulation Modeling Handbook: A Practical Approach, Christopher A. Chung, 2003.
- Object Oriented Simulation: A Modeling and Programming Perspective, Garrido, José M, 2009.
- Simulation, Modeling and Analysis, Averill M Law and W. David Kelton, "Tata McGraw Hill, Third Edition, 2003.
- Process Control: Modeling, Design and Simulation, Wayne Bequette W, Prentice Hall of India, 2003.

Course Code	Course Title	Credits		
SIPSCS421	Specialization: Cloud Computing -III	04		
	(Building Clouds and Services)			
Unit I: Cloud F	Reference Architectures and Security			
The NIST def	inition of Cloud Computing, Cloud Computing reference a	architecture,		
Cloud Computi	ng use cases, Cloud Computing standards. Cloud Computin	ng Security-		
Basic Terms	and Concepts, Threat Agents, Cloud Security Thre	ats. Cloud		
Security Mech	anisms, Encryption, Hashing, Digital Signature, P	ublic Key		
Infrastructure	(PKI), Identity and Access Management (IAM), Single Sign	i-On (SSO),		
Cloud-Based S	ecurity Groups, Hardened Virtual Server Images.			
Unit II: Cloud	Computing Mechanisms			
Cloud Infrastru	icture Mechanisms, Logical Network Perimeter, Virtual Se	erver, Cloud		
Storage Devi	ce, Cloud Usage Monitor, Resource Replication F	Ready-Made		
Environment.	Specialized Cloud Mechanisms, Automated Scaling	J Listener,		
Load Balancer	SLA Monitor, Pay-Per-Use Monitor, Audit Monito	r, Failover		
System, Hyper	visor, Resource Cluster, Multi-Device Broker, State M	lanagement		
Database. Clou	ud Management Mechanisms, Remote Administration	n System,		
Resource Mar	nagement System, SLA Management System, Billing M	lanagement		
System.				
Unit III: Cloud	Computing Architecture			
Fundamental C	loud Architectures, Workload Distribution Architecture, Resou	urce Pooling		
Architecture, D	ynamic Scalability Architecture, Elastic Resource Capacity A	Architecture,		
Service Load	Balancing Architecture, Cloud Bursting Architecture, E	lastic Disk		
Provisioning	Architecture, Redundant Storage Architecture. Advan	ced Cloud		
Architectures,	Hypervisor Clustering Architecture, Load Balanced Virt	tual Server		
Instances Arch	Instances Architecture, Non-Disruptive Service Relocation Architecture, Zero Downtime			
Architecture,	Cloud Balancing Architecture, Resource Reservation A	Architecture,		
Dynamic Failu	re Detection and Recovery Architecture, Bare-Metal	Provisioning		
Architecture,	Rapid Provisioning Architecture, Storage	Workload		

Management

Unit IV: Working with Clouds

Cloud Delivery Model Considerations, Cloud Delivery Models: The Cloud Provider Perspective, Building IaaS Environments, Equipping PaaS Environments, Optimizing SaaS Environments, Cloud Delivery Models: The Cloud Consumer Perspective. Cost Metrics and Pricing Models, Business Cost Metrics, Cloud Usage Cost Metrics, Cost Management Considerations. Service Quality Metrics and SLAs, Service Quality Metrics, Service Availability Metrics, Service Reliability Metrics, Service Performance Metrics, Service Scalability Metrics, Service Resiliency

Metrics.Text book:

- Cloud Computing Concepts, Technology & Architecture, Thomas Erl, Zaigham Mahmood, and Ricardo Puttini, Prentice Hall, 2013.
- Cloud Security A Comprehensive Guide to Secure Cloud Computing, Ronald L. Krutz, Russell Dean Vines, Wiley Publishing, Inc., 2010.
- Open Stack Cloud Computing Cookbook, Kevin Jackson, Cody Bunch, Egle Sigler, Packt Publishing, Third Edition, 2015.

- Tom Fifield, Diane Fleming, Anne Gentle, Lorin Hochstein, Jonathan Proulx, Everett Toews, and Joe, Topjian, OpenStack Operations Guide, O'Reilly Media, Inc, 2014.
- NIST Cloud Computing Standards Roadmap, Special Publication 500-291, Version 2, NIST, July 2013, http://www.nist.gov/itl/cloud/upload/NIST_SP-500-291_Version-2_2013_June18_FINAL.pdf
- https://www.openstack.org
- http://cloudstack.apache.org
- http://www.foss-cloud.org/en/wiki/FOSS-Cloud
- http://www.ubuntu.com/cloud/openstack/autopilot

Course Code	Course Title	Credits
SIPSCS422	Specialization: Cyber and Information Security	04
	(Cryptography and Crypt Analysis)	

Unit I: Introduction to Number Theory

Topics in Elementary Number Theory: O and notations, time estimates for doing arithmetic-divisibility and the Euclidean algorithm, Congruence: Definitions and properties, linear congruence, residue classes, Euler's phi function, Fermat's Little Theorem, Chinese Reminder Theorem, Applications to factoring, finite fields, quadratic residues and reciprocity: Quadratic residues, Legendre symbol, Jacobi Symbol. (proofs of the theorems are not expected to cover).

Unit II: Simple Cryptosystems

Shift Cipher, Substitution Cipher, Affine Cipher, Vigenère Cipher, Vermin Cipher, Hill Cipher, Permutation Cipher, Stream Cipher, Cryptanalysis of Affine Cipher, Substitution Cipher, Vigenère Cipher and Hill Cipher, Block Ciphers, Algorithm Modes, DES, Double DES, Triple DES, Meet-in-Middle Attack, AES, IDEA algorithm. Cryptographic Hash Functions: Hash Functions and Data Integrity, Security of Hash Functions, Secure Hash Algorithm, Message Authentication Code, Nested MACs, HMAC.

Unit III: RSA Cryptosystem

The RSA Algorithm, Primarily Testing, Legendre and Jacobi Symbols, The Solovay- Strassen Algorithm, The Miller-Rabin Algorithm, Factoring Algorithm: The pollard p-1 Algorithm, Dixon's Random Squares Algorithm, Attacks on RSA, The Rabin Cryptosystem. Public Key Cryptosystems: The idea of public key Cryptography, The Diffie-Hellman Key Agreement, ElGamal Cryptosystem, The Pollard Rho Discrete Logarithm Algorithm, Elliptic Curves, Knapsack problem.

Unit IV: Key Distribution and Key Agreement Scheme

Diffie-Hellman Key distribution and Key agreement scheme, Key Distribution Patterns, Mitchell-Piper Key distribution pattern, Station-to-station protocol, MTI Key Agreement scheme. Public-Key Infrastructure: What is PKI?, Secure Socket Layer, Certificates, Certificate Life cycle, Trust Models: Strict Hierarchy Model, Networked PKIs, The web browser Model, Pretty Good Privacy.

Text book:

- Discrete Mathematics and Its Applications, Kenneth H. Rosen, 7th Edition, McGraw Hill, 2012.
- Cryptography Theory and Practice, 3rd Edition, Douglas R. Stinson, 2005.

Reference:

- Network Security and Cryptography, Atul Kahate, McGraw Hill, 2003.
- Cryptography and Network Security: Principles and Practices, William Stalling, Fourth Edition, Prentice Hall, 2013.
- Introduction to Cryptography with coding theory, second edition, Wade Trappe, Lawrence C. Washington, Pearson, 2005.

Course Code	Course Title	Credits		
SIPSCS423	Specialization: Business Intelligence and Big Data	04		
	Analytics (Intelligent Data Analysis)			
Unit I: Clusteri	ng			
Distance/Simila	rity, Partitioning Algorithm: K-Means; K-Medoids, Partitionin	g Algorithm		
for large data	a set: CLARA; CLARANS, Hierarchical Algorithms: Ag	glomerative		
(AGNES); Div	(AGNES); Divisive (DIANA), Density based clustering: DBSCAN, Clustering i			
Non- Euclidean	Spaces, Clustering for Streams and Parallelism.			
Unit II: Classif	Unit II: Classification			
Challenges, Distance based Algorithm: K nearest Neighbors and kD-Trees, Rules and				
Trees based Classifiers, Information gain theory, Statistical based classifiers: Bayesian				

classification, Document classification, Bayesian Networks. Introduction to Support

Vector Machines, Evaluation: Confusion Matrix, Costs, Lift Curves, ROC Curves, Regression/model trees: CHAID (Chi Squared Automatic Interaction Detector). CART (Classification And Regression Tree).

Unit III: Dimensionality Reduction

Introduction to Eigen values and Eigen vectors of Symmetric Matrices, Principal-Component Analysis, Singular-Value Decomposition, CUR Decomposition.

Unit IV: Link Analysis And Recommendation Systems

Link analysis: PageRank, Efficient Computation of PageRank, Topic-Sensitive PageRank, Link Spam. Recommendation Systems: A Model for Recommendation Systems, Content-Based Recommendations, Collaborative Filtering, Dimensionality Reduction.

Text book:

- Mining of Massive Datasets, Anand Rajaraman and Jeffrey David Ullman, Cambridge University Press, 2012.
- Data Mining: Introductory and Advanced Topics, Margaret H. Dunham, Pearson, 2013.

- Big Data for Dummies, J. Hurwitz, et al., Wiley, 2013.
- Networks, Crowds, and Markets: Reasoning about a Highly Connected World, David Easley and Jon Kleinberg, Cambridge University Press, 2010.
- Lecture Notes in Data Mining, Berry, Browne, World Scientific, 2009.
- Data Mining: Concepts and Techniques third edition, Han and Kamber, Morgan Kaufmann, 2011.
- Data Mining Practical Machine Learning Tools and Techniques, Ian H.
 Witten, Eibe Frank, The Morgan Kaufmann Series in Data Management Systems, 2005.
- Big Data Analytics: From Strategic Planning to Enterprise Integration with Tools, Techniques, NoSQL and Graph, David Loshin, Morgan Kaufmann
 Publishers, 2013

Course Code	Course Title	Credits		
SIPSCS424	Specialization: Machine Learning -III	04		
	(Computational Intelligence)			
Unit I: Artificia	I Neural Networks			
The Artificial N Learning	leuron, Supervised Learning Neural Networks, Unsupervis	sed		
Neural Netwo	rks, Radial Basis Function Networks, Reinforcemen	t		
Learning, Perfo	rmance Issues.			
Introduction to Programming, I Evolution, Cul	Unit II: Evolutionary Computation Introduction to Evolutionary Computation, Genetic Algorithms, Genetic Programming, Evolutionary Programming, Evolution Strategies, Differential Evolution, Cultural Algorithms, Co-evolution.			
Particle Swarm Structures, Ba	utational Swarm Intelligence Optimization(PSO) - Basic Particle Swarm Optimization, So asic Variations and parameters, Single-Solution PSO. plications. Ant Algorithms- Ant Colony Optimization Me	Advanced		
Organization ar	nd Brood Care, Division of Labor, Advanced Topics and applie	cations.		

Unit IV: Artificial Immune systems, Fuzzy Systems and Rough Sets

Natural Immune System, Artificial Immune Models, Fuzzy Sets, Fuzzy Logic and Reasoning, Fuzzy Controllers, Rough Sets.

Text book:

• Computational Intelligence- An Introduction (Second Edition): Andries P. Engelbrecht, John Willey & Sons Publications (2007).

- Computational Intelligence And Feature Selection: Rough And Fuzzy Approaches, Richard Jensen Qiang Shen, IEEE Press Series On Computational Intelligence, A John Wiley & Sons, Inc., Publication, 2008.
- Computational Intelligence And Pattern Analysis In Biological Informatics, (Editors). Ujjwal Maulik, Sanghamitra Bandyopadhyay, Jason T. L.Wang, John Wiley & Sons, Inc, 2010.

- Neural Networks for Applied Sciences and Engineering: From Fundamentals to Complex Pattern Recognition 1st Edition, Sandhya Samarasinghe, Auerbach Publications, 2006.
- Introduction to Evolutionary Computing (Natural Computing Series) 2nd ed, A.E.
 Eiben , James E Smith, Springer; 2015.
- Swarm Intelligence, 1st Edition, Russell C. Eberhart, Yuhui Shi, James Kennedy, Morgan Kaufmann, 2001
- Artificial Immune System: Applications in Computer Security, Ying Tan, Wiley- IEEE Computer Society, 2016.
- Computational Intelligence and Feature Selection: Rough and Fuzzy Approaches 1st Edition, Richard Jensen, Qiang Shen, Wiley-IEEE Press, 2008

Cou	rse Code	Course Title	Credits
SIPS	SCSP41	Practical course on Simulation and modeling	02
Sr			
No		List of Practical Experiments	
1	Design a	nd develop agent based model by	
	• Cre	eating the agent population	
	• De	fining the agent behavior	
	• Ad	d a chart to visualize the model output.	
	[U	se a case scenario like grocery store, telephone call center e	tc for the
	pu	rpose].	
2	Design a	nd develop agent based model by	
	• Cre	eating the agent population	
	• De	fining the agent behavior	
	• Ad	ding a chart to visualize the model output	
	• Ad	ding word of mouth effect	
	• Co	nsidering product discards	

List of Practical Experiments for Semester –IV

	Considering delivery time				
	[Use a case scenario like restaurant].				
3	Design and develop agent based model by				
	Creating the agent population				
	 Defining the agent behavior 				
	 Adding a chart to visualize the model output 				
	5				
	Adding word of mouth effect				
	Considering product discards				
	Consider delivery time				
	Simulating agent impatience				
	 Comparing model runs with different parameter values 				
	[Use a scenario like market model]				
4	Design and develop System Dynamic model by				
	Creating a stock and flow diagram				
	Adding a plot to visualize dynamics				
	Parameter Variation				
	Calibration				
	[Use a case scenario like spread of contagious disease for the purpose]				
5	Design and develop a discrete-event model that will simulate process by:				
	Creating a simple model				
	Adding resources				
	Creating 3D animation				
	Modeling delivery				
	[Use a case situation like a company's manufacturing and shipping].				
6	Design and develop time-slice simulation for a scenario like 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 secur				
	checkpoint and then go to the waiting area. After boarding starts, each airline's				
	representatives check their passengers' tickets before they allow them to boa				

7	Verify and validate a model developed like bank model or manufacturing model				
8	Create defense model to stimulate aircraft behavior				
9	Stimulate the travelling sales man problem to compute the shortest path.				
10	Stimulate the Urban dynamics to address the scenarios like:				
	(a) The problem of public transport line				
	(b) To compute the time taken for train to enter the station				

Course Code		Course Title	Credits		
SIPSCSP421		Practical Course on Specialization: Cloud Computing			
		02 (Building Clouds and Services)			
Sr		List of Practical Experiments			
No					
1	Develop a private cloud using an open source technology.				
2	Develop a public cloud using an open source technology.				
3	Explore Service Offerings, Disk Offerings, Network Offerings and Templates.				
4	Explore Working of the following with Virtual Machines				
	•	VM Lifecycle			
	•	Creating VMs			
	•	Accessing VMs			
	•	Assigning VMs to Hosts			
5	Explore Working of the following with Virtual Machines				
	•	Changing the Service Offering for a VM			
	Using SSH Keys for Authentication				
6	Explore the working of the following: Storage Overview				
	•	Primary Storage			

	Secondary Storage						
7	Explore the working of the following: Storage Overview						
	Working With Volumes						
	Working with Volume Snapshots						
8	Explore managing the Cloud using following:						
	Tags to Organize Resources in the Cloud						
	Reporting CPU Sockets						
9	Explore managing the Cloud using following:						
	Changing the Database Configuration						
	File encryption type						
10	Explore managing the Cloud using following:						
	Administrator Alerts						
	Customizing the Network Domain Name						
Note							
Reco	ommended Open Source Technologies for completing practical:						
	FOSS-Cloud						
	Try Stack						
	Apache CloudStack						
	OpenStack						
	Canonical's OpenStack Autopilot						
	Recommended Configuration: Desktop PC Core I5 with minimum 250 GB Hard Drive and minimum 8 GB RAM						

Cou	rse Code	Course Title	Credits
SIPSCSP422		Practical Course on Specialization: Cyber &	02
		Information Security (Cryptography and Crypt	
		Analysis)	
Sr		List of Practical Experiments	
No			
1	Write a p	rogram to implement following:	
	•	Chinese Reminder Theorem	
	•	Fermat's Little Theorem	
2	Write a p	program to implement the (i) Affine Cipher (ii) Rail Fence Te	echnique (iii)
	Simple	Columnar Technique (iv) Vermin Cipher (v) Hill	Cipher to
	perform e	encryption and decryption.	
3	Write a and decry	program to implement the (i) RSA Algorithm to perform	n encryption
4	Write a	program to implement the (i) Miller-Rabin Algorithm (ii)	pollard p-1
	Algorithm	n to perform encryption and decryption.	
5	Write a p	program to implement the ElGamal Cryptosystem to generation	ate keys and
	perform e	encryption and decryption.	
6	Write a	program to implement the Diffie-Hellman Key Agreeme	nt algorithm
	to genera	ate symmetric keys.	
7	Write a p	rogram to implement the MD5 algorithm compute the message	ge digest.
8	Write a p	program to implement different processes of DES algorithm	like (i) Initial
	Permutat	ion process of DES algorithm, (ii) Generate Keys for DES a	lgorithm, (iii)
	S-Box su	bstitution for DES algorithm.	
9	Write a p	rogram to encrypt and decrypt text using IDEA algorithm.	
10	Write a p	rogram to implement HMAC signatures.	

Cours	se Code	Course Title	Credits					
SIPSCSP423		Practical Course on Specialization:	02					
		Business Intelligence & Big Data Analytics						
		(Intelligent Data Analysis)						
Sr		List of Practical Experiments	1					
No								
1	Pre-proce	ess the given data set and hence apply clustering techn	iques like					
	K- Means	s, K-Medoids. Interpret the result.						
		ss the given data set and hence apply partition clusterir s. Interpret the result	ng					
	•	ess the given data set and hence apply hierarchical algorated as a set and hence apply hierarchical algorated a	prithms and					
4	Pre-proce	ess the given data set and hence classify the resultant	data					
5		ess the given data set and hence classify the resultant atistical based classifiers. Interpret the result.	data set					
6	•	ess the given data set and hence classify the resultant port vector machine. Interpret the result.	data set					
7	Write a p	rogram to explain different functions of Principal Components	S.					
8	Write a p	rogram to explain CUR Decomposition technique.						
	Write a program to explain links to establish higher-order relationships Among entities in Link Analysis.							
	Write a program to implement step-by-step a Collaborative Filtering Recommender System.							
The ex	xperiment	ts may be done using software/ tools like R/Weka/Java etc.						
	xperiment							

Course Code		se Code Course Title					
PSCSP2024		Practical Course on Specialization:	02				
		Machine Intelligence					
Sr		List of Practical Experiments					
No							
1	Impleme	nt feed forward neural network for a given data.					
2	Impleme	nt Self Organizing Map neural network.					
3	Impleme	nt Radial Basis Function neural network with gradient desce	ent.				
4	Implemer	nt a basic genetic algorithm with selection, mutation and cro	ssover as				
	genetic o	perators.					
5	Impleme	nt evolution strategy algorithm.					
6	Impleme	nt general differential evolution algorithm.					
7	Impleme	nt gbest and lbest of PSO.					
8	Impleme	nt simple Ant colony optimization algorithm.					
9	Impleme	nt basic artificial immune system algorithm.					
10	Apply diff	erent defuzzification methods for centroid calculation of a g	iven fuzzy				
	rule base						
Note	e: The abov	ve practical experiments may use programming languages	ike C, Java,				
R et	с.						

Scheme of Examination for Theory Courses

There will be internal and external examination for the theory courses. The weightage of internal/external and scheme of examination will be as per common guidelines provided by the University for the PG courses in the faculty of Science.

Scheme of Examination for Practical Courses

There will not be any internal examination for practical courses.

External Examination for practical courses:

The evaluation of the external examination of practical course is given below:

Sr	Semester	Course	Particular		Particular		No of	Marks	Total
No		Code			questions	per	Marks		
						question			
			Laboratory e	xperiment					
	111	question		2	40	80			
	111	SIPSCS	Journal		-	10	10		
1		P5	Viva		-	10	10		
		Ma	arks for each cours	Se		100			
			Laboratory e	xperiment	2	25	50		
			question						
			Journal		-	10	10		
2	III	SIPSCSP6	Viva		-	10	10		
			viva on Project	Documen	tation	10			
			Proposal	Proposal Presentat		10	30		
				Viva		10			
			Total Marks	1	100				

Sr	Semester	Course	F	Particular		No of	Marks	Total	
No		Code				questions	per	Marks	
							question		
			Laborato	ry exper	iment				
			question			2	40	80	
1	IV	SIPSCSP7	Journal			-	10	10	
			Viva			-	10	10	
				Total Marks	tal Marks)	
						ty and	40		
			Intern- ship	Internship conduct	releva Docui	ance mentation	30	100	
2	N /				Prese	entation	30		
	IV	SIPSCSP8		Internship	Viva		50	50	
				Total Marks 150)		
					Qualit	ty and	40		
			Project	Project	releva	ance		100	
			Implom	conduct	Docui	mentation	30		
3	IV	SIPSCSP9	Implem entation		Prese	entation	30		
				Project viva	3		50	50	
				Total Marks)	

Guide lines for maintenance of journals:

A student should maintain a journal with at least six practical experiments for each part of the practical course. Certified journals need to be submitted at the time of the practical examination.

Guidelines for Project Proposal in Semester - III

- Student should take a topic related to the specialization he or she is planning to take in Semester-IV.
- Should have studied the related topics in the elective he or she has chosen in semester-II and semester- III
- A student is expected to devote at least 2 to 3 months of study as part of topic selection and its documentation.
- The student should be comfortable to implement the proposal in the semester IV.

Guidelines for Documentation of Project Proposal in Semester -III

Student is expected to make a project proposal documentation which should contain the following:

- **Title:** A suitable title giving the idea about what work is proposed.
- Introduction: An introduction to the topic of around 3-5 pages, giving proper back ground of the topic discussed.
- Related works: A detailed survey of the relevant works done by others in the domain. Student is expected to refer at least 5 research papers in addition to text books and web-links in the relevant topic. It may be around 7 to 10 pages.
- Objective: A detailed objective of the proposal is needed. It may be of 1 to 2 pages.

• **Methodology:** A proper and detailed procedure of how to solve the problem discussed. It shall contain the techniques, tools, software and data to be used. It shall be of around 3 to 5 pages.

The report may be of around 20 pages, which needs to be signed by the teacher in charge and head of the Department. Students should submit the signed project proposal documentation at the time of viva as part of the University examination.

Guidelines for internship in Semester - IV

- Internship should be of 2 to 3 months with 8 to 12 weeks duration.
- A student is expected to find internship by himself or herself. However, the institution should assist their students in getting internship in good organizations.
- The home institution cannot be taken as the place of internship.
- A student is expected to devote at least 300 hours physically at the organization.
- Internship can be on any topic covered in the syllabus mentioned in the syllabus, not restricted to the specialization.
- Internship can be done, in one of the following, but not restricted to, types of organizations:
 - Software development firms
 - Hardware/ manufacturing firms
 - o Any small scale industries, service providers like banks
 - o Clinics/ NGOs/professional institutions like that of CA, Advocate etc
 - Civic Depts like Ward office/post office/police station/ punchayat.
 - Research Centres/ University Depts/ College as research Assistant for research projects or similar capacities.

Guidelines for making Internship Report in Semester –IV

A student is expected to make a report based on the internship he or she has done in an organization. It should contain the following:

• Certificate: A certificate in the prescribed Performa (given in appendix 1) from

the organization where the internship done.

• Evaluation form: The form filled by the supervisor or to whom the intern was reporting, in the prescribed Performa (given in appendix 2).

- **Title:** A suitable title giving the idea about what work the student has performed during the internship.
- **Description of the organization:** A small description of 1 to 2 pages on the organization where the student has interned
- Description about the activities done by the section where the intern has worked: A description of 2 to 4 pages about the section or cell of the organization where the intern actually worked. This should give an idea about the type of activity a new employee is expected to do in that section of the organization.
- Description of work allotted and actually done by the intern: A detailed description of the work allotted and actual work performed by the intern during the internship period. Intern may give a weekly report of the work by him or her if needed. It shall be of around 7 to 10 pages.
- **Self assessment:** A self assessment by the intern on what he or she has learnt during the internship period. It shall contain both technical as well as inter personal skills learned in the process. It shall be of around 2 to 3 pages.

The internship report may be around 15 pages and this needs to be submitted to the external examiner at the time of examination.

Guidelines for Research Implementation in Semester - IV

- Student should continue with topic proposed and evaluated at the semester III.
- The topic has to be related with the specialization he or she has chosen in the semester – IV.
- A student is expected to devote at least 3 to 4 months of efforts for the implementation.
- Student should submit a detailed project implementation report at the time of viva.

Guidelines for Documentation of Project Proposal in Semester –IV

A Student should submit project implementation report with following details:

- **Title:** Title of the project (Same as the one proposed and evaluated at the semester II examination).
- Implementation details: A description of how the project has been implemented. It shall be of 2 to 4 pages.
- Experimental set up and results: A detailed explanation on how experiments were conducted, what software used and the results obtained. Details like screen shots, tables and graphs can come here. It shall be of 6 to 10 pages.
- Analysis of the results: A description on what the results means 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.
- Conclusion: A conclusion of the project performed in terms of its outcome (May be half a page).
- **Future enhancement:** A small description on what enhancement can be done when more time and resources are available (May be half a page).
- **Program code:** The program code may be given as appendix.

The report may be of around 20 pages (excluding program code), which needs to be signed by the teacher in charge and head of the Department. Student should submit the signed project implementation report along with evaluated copy of the project proposal documentation (of semester –III) at the time of Project evaluation and viva as part of the examination.

Appendix 1

(Proforma for the certificate for internship in official letter head)

This	is	to	certify	that	Mr/Ms							_	of
			(College/	Institution	worked	as	an	intern	as	part	of	her
MSc	course	in	Computer	Scienc	e of SIES	College	of	Arts	, Scien	ice 8	& Cor	nme	rce,
Sion(West) (Aut	onomous) .	The pa	articulars of	internshi	p ar	e giv	en belo	w:			
Intern	nship sta	artir	ng date:										
Intern	nship er	ndin	g date:										
Actua	al numb	er c	of days worl	ked:									
Tenta	ative nu	mbe	er of hours	worked		Hours							
Broad	d area c	of w	ork:										
A sm	all desc	ript	ion of work	done b	y the intern	during th	ne p	eriod	:				
Signa	ature:												
Name	e:												
Desig	nation:												
Conta	act num	ber	:										
Email	l:												

(seal of the organization)

Appendix 2

(Proforma for the Evaluation of the intern by the supervisor/to whom the intern was

reporting in the organization)

Professional Evaluation of intern

Name of intern:_____

College/institution:_____

[Note: Give a score in the 1-5 scale by putting $\sqrt{}$ in the respective cells]

Sr	Particular	Excellent	Very	Good	Moderate	Satisfactory
No			Good			
1	Attendance					
2	Punctuality					
3	Adaptability					
4	Ability to shoulder responsibility					
5	Ability to work in a team					
6	Written and oral communication skills					
7	Problem solving skills					
8	Ability to grasp new concepts					
9	Ability to complete task					
10	Quality of work done					

Comments:

Signature: Name:

Designation:

Contact number:

Email:

(seal of the organization)