

# **School of Computer Science and Engineering**

# CURRICULUM AND SYLLABI (2019-2020)

M.Tech (CSE) - Specialisation in Big Data Analytics



#### VISION STATEMENT OF VELLORE INSTITUTE OF TECHNOLOGY

Transforming life through excellence in education and research.

# MISSION STATEMENT OF VELLORE INSTITUTE OF TECHNOLOGY

**World class Education**: Excellence in education, grounded in ethics and critical thinking, for improvement of life.

**Cutting edge Research**: An innovation ecosystem to extend knowledge and solve critical problems.

Impactful People: Happy, accountable, caring and effective workforce and students.

**Rewarding Co-creations**: Active collaboration with national & international industries & universities for productivity and economic development.

Service to Society: Service to the region and world through knowledge and compassion.

# VISION STATEMENT OF THE SCHOOL OF COMPUTER SCIENCE AND ENGINEERING

To be a world-renowned centre of education, research and service in computing and allied domains.

# MISSION STATEMENT OF THE SCHOOL OF COMPUTER SCIENCE AND ENGINEERING

- To offer computing education programs with the goal that the students become technically competent and develop lifelong learning skill.
- To undertake path-breaking research that creates new computing technologies and solutions for industry and society at large.
- To foster vibrant outreach programs for industry, research organizations, academia and society.



# School of Computer Science and Engineering

M.Tech (CSE) - Specialisation in Big Data Analytics

# PROGRAMME EDUCATIONAL OBJECTIVES (PEOs)

- 1. Graduates will be engineering professionals who will engage in technology development and deployment with social awareness and responsibility.
- 2. Graduates will function as successful practicing engineer / researcher / teacher / entrepreneur in the chosen domain of study.
- 3. Graduates will have holistic approach addressing technological, societal, economic and sustainability dimensions of problems and contribute to economic growth of the country.



# M. Tech Computer Science and Engineering Specialization in Big Data Analytics

## **PROGRAMME OUTCOMES (POs)**

- PO\_1 Having an ability to apply mathematics and science in engineering applications
- PO\_2 Having an ability to design a component or a product applying all the relevant standards and with realistic constraints
- PO\_3 Having an ability to design and conduct experiments, as well as to analyze and interpret data
- PO\_4 Having an ability to use techniques, skills and modern engineering tools necessary for engineering practice
- PO\_5 Having problem solving ability- solving social issues and engineering problems
- PO\_6 Having adaptive thinking and adaptability
- PO\_7 Having a clear understanding of professional and ethical responsibility
- PO\_8 Having a good cognitive load management [discriminate and filter the available data] skills



# School of Computer Science and Engineering

M.Tech (CSE) - Specialisation in Big Data Analytics

### PROGRAMME SPECIFIC OUTCOMES (PSOs)

- 1. Ability to design and develop computer programs/computer-based systems in the advanced level of areas including algorithms design and analysis, networking, operating systems design, etc.
- 2. Ability to apply the advanced concepts of Big Data that pave the way to create a platform to gain analytical skills which impacts business decisions and strategies.
- 3. Ability to bring out the capabilities for research and development in contemporary issues and to exhibit the outcomes as technical report.



# M. Tech Computer Science and Engineering Specialization in Big Data Analytics

# **CREDIT STRUCTURE**

## **Category-wise Credit distribution**

Category	Credits
University Core (UC)	27
Programme Core (PC)	18
Programme Elective (PE)	19
University Elective (UE)	06
Bridge Course (BC)	-
Total Credits	70



### CURRICULUM M.Tech.-CSE (Spl. in Big Data Analytics) - (2019)

Programme Core	Programme Elective	University Core	University Elective	Total Credits
18	19	27	6	70

Course Code	Course Title	Course Type	L	Т	Р	J	С	
	PROGRAMME CORE							
CSE5001	Algorithms: Design and Implementation	ETL	2	0	2	0	3	
CSE5003	Database Systems: Design and Implementation	ETLP	2	0	2	4	4	
CSE5007	Exploratory Data Analysis	ETP	2	0	0	4	3	
CSE6001	Bigdata Frameworks	ETLP	2	0	2	4	4	
CSE6005	Machine Learning	ETLP	2	0	2	4	4	
Course Code	Course Title	Course Type	L	Т	Р	J	С	
	PROGRAMME ELECTIV	<b>/E</b>						
CSE5002	Operating Systems and Virtualization	ETL	2	0	2	0	3	
CSE6006	NoSQL Databases	ETLP	2	0	2	4	4	
CSE6014	Programming for Data Science	LO	0	0	4	0	2	
CSE6016	Information Visualization	ETLP	2	0	2	4	4	
CSE6017	Mining Massive Data	ETLP	2	0	2	4	4	
CSE6018	Streaming Data Analytics	ETLP	2	0	2	4	4	
CSE6019	Text, Web and Social Media Analytic	ETP	3	0	0	4	4	
CSE6020	Big Data Technologies	ETLP	2	0	2	4	4	
CSE6021	Domain Specific Predictive Analytics	ETP	3	0	0	4	4	
CSE6022	Soft Computing	ETP	3	0	0	4	4	
CSE6023	Cloud Computing Fundamentals	ETLP	2	0	2	4	4	
CSE6025	Analytics of Things	ETP	3	0	0	4	4	
CSE6041	Blockchain Technology	ETP	2	0	0	4	3	
CSE6042	Deep Learning	ETL	2	0	2	0	3	
CSE6043	Image and Video Analytics	ETP	2	0	0	4	3	
CSE6046	Network Science and Applications	ETL	3	0	2	0	4	
Course Code	Course Title	Course Type	L	Т	Р	J	С	
	UNIVERSITY CORE							
CSE6099	Masters Thesis	PJT	0	0	0	0	16	
MAT6001	Advanced Statistical Methods	ETL	2	0	2	0	3	
SET5001	Science, Engineering and Technology Project - I	PJT	0	0	0	0	2	
SET5002	Science, Engineering and Technology Project - II	PJT	0	0	0	0	2	
EFL5097	English and Foreign Language	CDB	0	0	0	0	2	
ENG5001 - Fundame	ntals of Communication Skills - LO							
ENG5002 - Professio	nal and Communication Skills - LO							
	FRE5001 - Francais fonctionnel – TH							
GER5001 - Deutsch f		T						
STS6777	Soft Skills M.Tech.	CDB	0	0	0	0	2	
STS5001 - Essentials	s of Business Etiquettes - SS							

STS5001 - Essentials of Business Etiquette and Problem Solving - SS

STS5002 - Preparing for Industry – SS

STS5102 - Programming and Problem Solving Skills - SS



### CURRICULUM M.Tech.-CSE (Spl. in Big Data Analytics) - (2019)

Course Code	Course Title	Course Type	L	Т	Р	J	С	
Course Code	Course Title	Course Type	L	Т	Р	J	С	
BRIDGE COURSE								
Course Code	Course Title	Course Type	L	т	Р	J	С	
NON CREDIT COURSE								

CSE5001	ALGORITHMS: DESIGN AND IMPLEMENTATION					J	C
			2	0	2	0	3
<b>Pre-requisite</b>	NIL	Syllabus version				s version	
							1.0

- 1. To focus on the design of algorithms in various domains
- 2. To provide a foundation for designing efficient algorithms.
- 3. To provide familiarity with main thrusts of working algorithms-sufficient to gives context for formulating and seeking known solutions to an algorithmic problem.

#### **Expected Course Outcome:**

- 1. Solve a problem using Algorithms and design techniques
- 2. Solve complexities of problems in various domains
- 3. Implement algorithm, compare their performance characteristics, and estimate their potential effectiveness in applications
- 4. Solve optimization problems using simplex algorithm
- 5. Designing approximate algorithms for graph theoretical problems
- 6. Application of appropriate search algorithms for graphs and trees
- 7. Application of computational geometry method on optimization problems

Module:1	Introduction	5 hours
_	sign techniques: Divide and Conquer, Brute force, Greedy, I symptotic notation, recurrence relations)	Dynamic Programming. Time
Module:2	Network Flows	5 hours
	ows, Min-cost Flows, Max-Flow Min-Cut Theorem, Cycle C me Analysis, Minimum Cuts without Flows	Canceling Algorithms, Strongly
Module:3	Tractable and Intractable Problems	4 hours
Class complex	ity: P, NP, NP-Hard, NP-Complete Approximation Algorithm	ms
Module:4 Limits to Appr	Approximation Algorithms oximability, Vertex Cover problem, Set cover problem, Euc	4 hours
Module:5	Search Algorithms for Graphs and Trees	4 hours
Overview of fu	undamental algorithms, Dijkstra's algorithm, A*search algor	ithm
Module:6	Computational Geometry	4 hours
Line Segments	, Convex hull finding algorithms	
Module:7	Linear Programming	2 hours

Modu	le:8 Recent Trends	2 hours
	Total Lecture hours:	30 hours
Text I	Book(s)	
D. C	D. I.	
Refere	1. Cormen, Leiserson, Rivest and Stein, Introduction to	Algorithms, 3rd edition, McGraw-
	<ol> <li>Hill, 2009.</li> <li>J.Kleinberg and E.Tardos. Algorithm Design, Pearson</li> <li>E.Horowitz, S.Sahni, S.Rajasekaran, Fundamentals of 2nd edition, Universities Press, 2011.</li> <li>Ravindra K.Ahuja, Thomas L. Magnanti, and James B Algorithms, and Applications, Pearson Education, 20</li> <li>George T. Heineman, Gary Pollice, Stanley Selkow Media, 2nd edition, 2016.</li> </ol>	f Computer Algorithms,  3. Orlin, Network Flows: Theory,  14.
Mode	of Evaluation: CAT / Assignment / Quiz / FAT / Project / S	Seminar
	f Challenging Experiments (Indicative)	
1.	Implementation of algorithms for problems that can be solved by of the following strategies: Divide and Conquer, Brute force, Greed Programming.	
2.	Implementation of Ford Fulkerson method, Edmonds-Karp algorifinding maximum flow in a flow network and applying them for typical problems such as railway network flow, maximum be matching	solving
3.	Implementation of Dinics strongly polynomial algorithm for commaximum flow in a flow network and applying it for solving typic	
4.	Implementation of push-relabel algorithm of Goldberg and Tarjan for finding maximum flow in a flow network and applying typical problems	2 hours
5.	Applying linear programming for solving maximum flow problem	n 2 Hours
6.	Applying network flow algorithms for baseball elimination and aischeduling	irline 2 Hours
7.	Given a flow network G=(V,E,s,t), where V is the vertex set, E is and t are source and destination. An edge of the flow network is a decrease in the flow over that edge results in a decrease in the to flow network. An edge of the flow network is called a bottlen increase in the flow over that edge results in an increase in the to flow network. Assume that you are using to compute the maxim network.	called critical if otal flow of the neck edge if an otal flow of the
	<ul><li>(a) Write a program (any language) to identify all the critic</li><li>(b) Write a program (any language) to identify all bottlen in the network.</li></ul>	
8.	Implementation of solution techniques for the minimum-cost flow	2 hours

	Problem				
9.	programming problem in constrain to f the problem the solution of the followard programming language. A and tables. Processing of A chair requires 2 hours or requires 5 hours on maching of time per day available	two diment, into a plant, into a plant, into a plant, into a plant manufacture, these products and machine M1 and into machine	anar region. blem. Implurer of furni ucts is done M1and 61 no time on 1 ne M1and30 r and a table	upute the solution of a linear ralgorithm should convert each. Use that algorithm to compute ement your algorithm in any ture makes two products: chairs to on two machines M1 and M2. Thours on machine M2. A table machine M2. There are 16 hours of hours on machine M2. Profits the are Rs.1and Rs.5 respectively.	2 hours
10.	Implementation of algorith TSP	hms for the	e vertex cov	er problem, set cover problem,	2 hours
11.	Implementation of search algorithms, Dijkstra's algo		hms for g	graphs and trees: fundamental	2 hours
12.	Forest officials have tran purpose. You are allow	ers by a fence of shortest length. Suggest an algorithm for the nformation required for your programming language (using	3 hours		
13.	13. A simple polygon is defined as a flat shape consisting of straight non-intersecting line segments or sides that are joined pairwise tofromaclosedpath.Letp1,p2,,pn be a set of points in the two dimensional plane. (a) Write a program to find the simple polygon of P. (b) Write a program (linear time) to convert that the simple polygon of P to a Convex Hull.				
	Tota		tory Hour		30 hours
D		N	Mode of as		
Keco	mmended by Board of Studies			13.05.2016	
Apj	proved by Academic Council	41	Date	17.06.2016	

CSE5003	DATABASE SYSTEMS: DESIGN AND IMPLEMENTATION	L	T	P	J	C
		2	0	2	4	4
Pre-requisite	NIL		Syllabu		us	
				ve	rsic	n
					1	0.

- 1. To emphasize the underlying principles of Relational Database Management System.
- 2. To model and design advanced data models to handle threat issues and counter measures.
- 3. To implement and maintain the structured, semi-structured and unstructured data in an efficient database system using emerging trends.

#### **Expected Course Outcome:**

- 1. Design and implement database depending on the business requirements and considering various design issues.
- 2. Select and construct appropriate parallel and distributed database architecture and formulate the cost of queries accordingly.
- 3. Understand the requirements of data and transaction management in mobile and spatial database and differentiate those with RDBMS.
- 4. Categorize and design the structured, semi-structured and unstructured databases.
- 5. Characterize the database threats and its counter measures.
- 6. Review cloud, streaming and graph databases.
- 7. Comprehend, design and query the database management system.

Module:1	Relational Model	6 hours
Database System optimization – Tran	Architecture–EER Modeling-Indexing–Normalisaction Processing	malization–Query processing and
Module:2	Parallel Databases	4 hours
Architecture, Data pa	rtitioning strategy, Interquery and Intraquery Parall	elism –Parallel Query Optimization
Module:3	Distributed Databases	5 hours
	outed Database Architecture –Fragmentation buted Transactions Processing	Replication- Distributed Query
Module:4	Spatial and Mobile Databases	3 hours
Spatial databases-Tyjin MDS	pe of spatial data–Indexing in spatial databases, Mo	bile Databases-Transaction Model
Module:5	Semi-Structured Databases	4 hours
Semi Structured data	bases – XML –Schema-DTD- XPath- XQuery, Ser	mantic Web –RDF–RDFS
Module:6	Database Security	3 hours

			rity Issues–Security Models–Different h these problems	Threats to datab	ases-
Mod	dule:7	Eme	rging Technologies		3 hours
			Databases - Graph Databases-New S	SQL	<u> </u>
			•		
Mod	dule:8	Recent Tren	nds		2 hours
			<b>Total Lecture hours:</b>	30 hours	
Tex	t Book(s)				
	2.	McGraw Hill, Ramez Elmas edition, Addis	tz, Hank Korth, and S.Sudarshan,"Datab 2010. Sri B.Navathe: "Fundamentals of data son Wesley,2014	·	
Ref	erence Bo				
		LSingh, "Datal erson education	base Systems: Concepts, Design A., 2011.	pplications", 2nd	d edition,
			ny Ayers, Liam R. E. Quin: "Begin: a Edition, 2012.	ning XML", Wi	ley India
	Ap		nolly and Carolyn Begg "Databas gn, Implementation, and Managemo	•	
Mod	lo of Evol	otion: CAT / A	Assignment / Quiz / FAT / Project / S	ominor	
			nents (Indicative)	emmai	
1.	Model an		into ER/EER Model using any tool ERI	D Plus, ER	3 hours
2.	Table cre	pplications with tion with constr d complex querion	aints, alter schema, insert values, aggreg	gate functions,	3 hours
	PLSQL-F	ROCEDURES,	CURSORS, FUNCTIONS, TRIGGERS		
3.		-	ase based on the type of query an uery with/without parallelism.	d compares the	3 hours
4.			and validate it against an XML Schema the contents of the database.	a/DTD. Use	3 hours
5.	in XML,I For each ,which of may hav players v	oTD and Xquery game, we want he was playing been penaltied were shown yell check your solu	which the results of football games are to to be able to represent the two team at home, which players scored goals(es) and the time when each was scordlow or red cards. You might use so attions with the online demo of the Zoats	s involved some of which red, and which ome attributes.	3 hours
6.	•		n and parallel sort algorithms to get marl and publish10 ranks for each discipline.	ks from different	2 hours

7.	Create a distributed database scenario, insequery the database.	rt values, fragme	ent the database	and			
8.	Consider a schema that contains the following table with the key underlined: Employee (Eno, Ename, Desg, Dno). Assume that we horizontally fragment the table as follows: Employee1(Eno;Ename;Desg;Dno), where 1;= Dno ;=10 Employee2(Eno;Ename; Desg; Dno), where 11 ;= Dno ;=20 Employee3 (Eno;Ename; Desg;Dno),where 21;=Dno;=30 In addition, assume we have 4 sites that contain the following fragments: Site1 has Employee1 Site2 has Employee2 Site3 has Employee2 and Employee3 Site4 has Employee1 Implement atleast 5 suitable queries on Employee fragments. Add relations to the database as per your requirements.						
9.	Download a spatial dataset based on an information) from Quantum GIS and import Query and view the database.				2 hours		
10.	To investigation of some spatial analysis techniques using Toxic Release Inventory (www.epa.gov/triexplorer/) data for Massachusetts from the Environmental Protection Agency (EPA),which indicate the magnitude of the releases of toxic core chemicals into land, water and air at a site in the state. Note that these TRI locations were geo coded from a list of addresses provided by the EPA						
11.	Use sample datasets from health care doma	in, Visualize and	l interpret the re	esults	3 hours		
12. Import the Hubway data intoNeo4jandconfigureNeo4j.Then, answer the following questions using the Cypher Query Language: a) List top 10 stations with most outbound trips (Show station name and number of trips) b) List top 10 stations with most inbound trips (Show station name and number of trips) c) List top 5 routes with most trips (Show starting station name, ending station name and number of trips)  (4) List the hour number (for example 13 means 1pm -2pm) and number of trips which start from the station" B.U.Central"  d)List the hour number(forexample13means1pm-2pm)and number of trips which end at the station "B.U. Central"					2 hours		
	Total Laboratory Hours 30 hou						
	le of assessment: Project/Activity			L			
	Recommended by Board of Studies 13.05.2016						
App	proved by Academic Council	41	Date	17.06.20	016		

				ITPIC
CSE5007		Exploratory Data Analy	sis	2 0 0 4 3
Pre-requisi	te	Nil		Syllabus version
				1.0
Course Obj	jectives	:		
2.It covers e through stat 3.Supports t Visualizatio	essentia istical r to Sumi on		nultivariate data es, data science a	by summarizing it nd Data
<ol> <li>Summarize</li> <li>Identify the</li> <li>Choose approximately</li> </ol>	ze the d he outli ppropri	data in the real world data sets by choosing a lata using basic statistics. Visualize the data ers if any in the data set. ate feature selection and dimensionality redu andling multi-dimensional data	using basic grapl	
		-	T	
Module:1		luction To Exploratory Data Analysis		3 hours
•		cycle, Exploratory Data Analysis (EDA)— Def	finition, Motivatio	on, Steps in data
Module:2	_	ocessing-Traditional Methods and num Likelihood Estimation		4 hours
		ng data, Traditional methods for dealing with dissing data handling, Improving the accuracy of	-	Maximum Likelihood
Module:3	Prepr	ocessing Bayesian Estimation		4 hours
		sian Estimation ,Multiple Imputation-Imputation ultiple Imputation, Models for Missing Notation		and Pooling Phase,
Module:4	Doto	Summarization & Visualization		4 hours
		ration, 1-D Statistical data analysis, 2-D Statistic	  cal data Analysis,	
Module:5		Outlier Analysis		3 hours
		ne Value Analysis, Clustering based, Distance E tection in Categorical Data	Based and Density	
Module:6		Feature Subset Selection		4 hours
Feature sele		gorithms: filter methods, wrapper methods and on, Relief, greedy selection, genetic algorithms f		ds, Forward selection
Module:7		Dimensionality Reduction		6 hours
	•	Dimensionality Reduction al Component Analysis(PCA), Kernel PCA, Car ensional scaling, Correspondence Analysis	nonical Correlation	6 hours n Analysis, Factor

**Recent Trends** 

	Total Lecture hours:	30 hours							
Tex	kt Book(s)								
Ref	ference Books								
1	Charu C. Aggarwal, "Data Mining The Text book", Sprin	ger, 2015.							
2	Craig K. Enders, "Applied Missing Data Analysis", The G	Guilford Press, 2010.							
3.	Inge Koch, "Analysis of Multivariate and High dimension	nal data", Cambridge University							
	Press, 2014.								
4.	Michael Jambu, "Exploratory and multivariate data analy	sis", Academic Press Inc. 1990.							
5.									
Mo	Mode of assessment:								
Rec	Recommended by Board of Studies 13-05-2016								
Ap	Approved by Academic Council No. 41 Date 17-06-2016								

CSE6001	BIG DATA FRAMEWORKS		L	T	P	J	C
			2	0	2	4	4
Pre-requisite	NIL			S	ylla	bus	version
		1				1.0	

- 1. To understand the need of Big Data, challenges and different analytical architectures
- 2.Installation and understanding of Hadoop Architecture and its ecosystems
- 3. Processing of Big Data with Advanced architectures like park.
- 4.Describe graphs and streaming data in Spark

#### **Expected Course Outcome:**

- 1. Discuss the challenges and their solutions in Big Data
- 2.Understand and work on Hadoop Framework and eco systems.
- 3. Explain and Analyse the Big Data using Map-reduce programming in Both Hadoop and Spark framework.
- 4. Demonstrate spark programming with different programming languages.
- 5. Demonstrate the graph algorithms and live streaming data in Spark
- 6. Analyse and implement different frame work tools by taking sample data sets.
- 7. Illustrate and implement the concepts by taking an application problem.

#### **Module:1** | **Introduction to Big Data**

3 hours

Data Storage and Analysis - Characteristics of Big Data - Big Data Analytics - Typical Analytical Architecture - Requirement for new analytical architecture - Challenges in Big Data Analytics - Need of big data frameworks

#### **Module:2** Hadoop Framework

6 hours

Hadoop – Requirement of Hadoop Framework - Design principle of Hadoop – Comparison with other system - Hadoop Components – Hadoop 1 vs Hadoop 2 – Hadoop Daemon's – HDFS Commands – Map Reduce Programming: I/O formats, Map side join, Reduce Side Join, Secondary sorting, Pipelining MapReduce jobs

#### **Module:3** | **Hadoop Ecosystem**

3 hours

Introduction to Hadoop ecosystem technologies: Serialization: AVRO, Co-ordination: Zookeeper, Databases: HBase, Hive, Scripting language: Pig, Streaming: Flink, Storm.

#### Module:4 | Spark Framework

4 hours

Overview of Spark – Hadoop vs Spark – Cluster Design – Cluster Management – performance, Application Programming interface (API): Spark Context, Resilient Distributed Datasets, Creating RDD, RDD Operations, Saving RDD - Lazy Operation – Spark Jobs.

Module:5	Data Analysis with	Spark Shell			4 hours
Writing Sp	ark Application - Spark Pro	gramming in Sca	ala, Pyt	hon, R, Jav	a - Application Execution.
Module:6	Spark SQL and Graph	X			5 hours
SQL Conte			es – usi	ng SQL – (	GraphX overview – Creating
Module:7	Spark Streaming				3 hours
Overview -	- Errors and Recovery – Str	eaming Source -	Stream	ning live da	ta with spark
Module:8	Recent Trends				2 hours
		Γotal Lecture ho	ours:	30 hours	
Reference	Books				
	1. Mike Frampton, "Maste	ring Apache Spark	x", Pack	t Publishing,	,2015.
	2. TomWhite,"Hadoop:Th	eDefinitiveGuide'	,O'Reill	y,4thEdition	n,2015.
	3. Nick Pentreath, Machin	e Learning with Sp	park, Pa	ckt Publishir	ng, 2015.
	4. Mohammed Guller, Big	Data Analytics wi	ith Sparl	k, Apress,20	15
	5. Donald Miner, Adam	Shook, "Map Re	duce D	esign Patte	rn", O'Reilly, 2012
	valuation: CAT / Assignme		Project	t / Seminar	
	allenging Experiments (In	dicative)			4 hours
	Commends Map Reduce Prog	gram to show the n	eed of C	Combiner	4 hours
-	educe I/O Formats-Text, ko Multiline	ey-value Map Re	duce I/0	O Formats -	_ 5 hours
3. Seque	nce file Input/Output Forma	ats Secondary son	rting		5 hours
	outed Cache & Map Side Jo			_	8 hours
	ng a Spark Application W ulating RDD	ord count in H	adoop a	and Spark	
Implen	d Indexing in Spark Sequentation of Matrix algorithm ag Spark Streaming application	ns in Spark Spar			
7.7.0			tal Lab	oratory H	ours 30 hours
	ssessment: <i>Project/Activity</i> nded by Board of	13.05.2016			
<b>Studies</b>	iucu by Doaru oi	13.03.2010			
	by Academic Council	No. 41	Date	17.06.20	16

CSE6005		MACHIN	E LI	EA]	RNI	NG					L	T	P	J	C
											2	0	2	4	4
Pre-requisit	e NIL										Sy	lla	bus	ver	sion
Course Obj	ectives:														1.0
·	cquire theoretica	Knowledge	on se	ettin	ng hvi	pothe	esis fo	or patt	ern r	eco	niti	ion			
	pply suitable ma	_				_		_					knov	wled	lge
from		•	U		1				U		U				υ
	valuate the perfo	rmance of alg	orith	ıms	and t	to pro	ovide	soluti	ion f	or v	ario	us 1	eal-	· wo	rld
appl	ications														
<b>Expected C</b>	ourse Outcome:														
	Recognize the ch	nracteristics o	f Ma	ichii	ne Le	earnii	ng tec	chniqu	ies th	nat e	nab	le to	o so	lveı	eal
2. F	Recognize the ch	racteristics of	f mac	chin	ne lea	ırninş	g stra	tegies							
3. A	Apply various su	ervised learn	ing n	netł	hods	to ap	propr	riate p	roble	ems					
	dentify and integearning	rate more tha	an on	ne te	echni	iques	s to er	nhanc	e the	per	fori	nan	ice (	of	
5. (	Create probabilis	ic and unsupe	ervise	ed le	earni	ng m	odels	for h	andli	ing ı	ınkr	ow	n pa	atter	'n
6. A	Analyze the co-o	currence of d	ata to	o fii	nd in	teres	ting f	reque	nt pa	tteri	ıs				
Module:1	INTRODUCT LEARNING	ON TO	]	MA	ACH	INE								3ho	ours
	, Examples of V te and Infinite H		_		_		-				s, V	ers.	ion		
Module:2	Supervised	Learning												9 h	ours
Multiple Lin	rees: ID3, Class near Regression, Perceptron, Supposebbours	Logistic Reg	gressi	ion,	, Net	ıral 1	Netwo	orks:	Intro	duc	tion	, Pe	erce	ptro	n,
Module:3	Ensemble Lear	ning												3 ho	ours
	oination Scheme , Boosting: Adab			orre	ecting	g Out	put C	Codes,	Bagg	ging	: Ra	ndo	om		
Module:4	Unsupervi	ed Learning												5ho	ours
	o clustering, Hiera expectation Maxim	rchical: AGNE	ES, DI					K-mea	ns clu	uster	ing,	K-N	Mod	e	
Module:5	Probabilist	ic Learning												3 h	ours
Bayesian Le	arning, Bayes O	otimal Classif	ier, N	Vaiv	ve Ba	yes (	Classi	ifier, I	Bayes	sian	Bel	ief	Net	wor	ks
Module:6	Learning Asso	ciation Rules	<u> </u>											3ho	ours
Mining Freq based Decisi	uent Patterns - b on Trees	asic concepts	-Apri	riori	i algo	rithn	n, FP	- Grov	wth a	lgor	ithn	n, A	Asso	ciat	ion-

Module:7

**Machine Learning in Practice** 

2 hours

	gn, Analysis and Evaluation of Machine Learning Experiments, Other Issue alanced data sets	es: Handling
11110	andreed data sets	
Mod	lule:8 Recent Trends in Big Data Analytics	2 hours
	Total Lecture hours: 30 hours	
Tex	t Book(s)	
Ref	erence Books	
	<ol> <li>Ethem Alpaydin, "Introduction to Machine Learning", MIT Press, Pr. Third Edition2014.</li> <li>Mehryar Mohri, Afshin Rostamizadeh, Ameet Talwalkar, "Foundati Learning", MIT Press, 2012.</li> <li>Tom Mitchell, "Machine Learning", McGraw Hill, 3<sup>rd</sup> Edition, 1997.</li> <li>Charu C. Aggarwal, "Data Classification Algorithms and Application Charu C. Aggarwal, "DATA CLUSTERING Algorithms and Application 2014.</li> <li>Kevin P. Murphy "Machine Learning: A Probabilistic Perspective", Jiawei Hanand Micheline Kambers and Jian Pei, "Data Mini Concepts and Techniques", 3rd edition, Morgan Kaufm</li> </ol>	ions of Machine
	Publications, 2012.	
Mod	le of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar	
List	of Challenging Experiments (Indicative)	
1.	Implement Decision Tree learning	2 hours
2.	Implement Logistic Regression	2 hours
3.	Implement classification using Multilayer perceptron	2 hours
4.	Implement classification using SVM	2 hours
5.	Implement Adaboost	2 hours
6.	Implement Bagging using Random Forests	2 hours
7.	Implement K-means Clustering to Find Natural Patterns in Data	2 hours
8.	Implement Hierarchical clustering	2 hours
9.	Implement K-mode clustering	2 hours
10.	Implement Association Rule Mining using FP Growth	2 hours
11.	Classification based on association rules	2 hours
12.	Implement Gaussian Mixture Model Using the Expectation Maximization	2 hours
13.	Evaluating ML algorithm with balanced and unbalanced datasets	2 hours
14.	Comparison of Machine Learning algorithms	2 hours
15.	Implement k-nearest neighbours algorithm	2 hours

	T	otal Labor	atory Hours	30 hours
Mode of assessment: Project/Activity				
<b>Recommended by Board of Studies</b>	13.05.2016			
Approved by Academic Council	No. 41	Date	17.06.2016	

CSE5002		OPERATING SYSTEMS AND VIRTUAL	IZATION	L T P J C
				2 0 2 0 3
Pre-requisi	te	Nil		Syllabus version
				1.
Course Ob	jectives:			
1. To introdu	ce Virtua	lization, operating systems fundamental concepts		
		write programs that interact with operating system	m components su	ch as processes, thread
memory duri	ng concu	rrent execution.		
		s and knowledge necessary to implement, provisio	ning and adminis	ster server and desktop
virtualization				
Expected C				
		ne course, the students will be able to		
		tem layers and kernel architectures.		
		niques for process management.		
		ddress translation mechanism.		
		eading and synchronization.		
		ods of virtualization and perform desktop and serv		
		eight virtual machines with dockers and containers related to the simulations of operating systems and		nconts
7. Develop p	rograms i	erated to the simulations of operating systems and	VII tualization coi	ncepts.
Module:1	INTD	DDUCTION		2 hour
		uter system architecture a layered view with interf	aces Glenford M	
		ernels Layered architecture of operating system an		
Tryona wine	IOWS TO K	erners Layered are intecture or operating system an	d core ranctional	11313.
Module:2	PROC	ESS		5 hour
		Operations, States, Context switching, Data Struct	res (Process Cor	
		Iulti-Level Feedback Queue, Multi-processor Sche		
			<u> </u>	
Module:3	MEM(	ORY		4 hour
Introduction,	Address	Spaces, Memory API, Address Translation, Pagin	g - Faster Transla	ntions (TLB), Smaller
Tables. Virtu	al Memo	ry System in x86.	-	
-				
Module:4	ll .	URRENCY		6 hour
		Models, Thread API, Building Evaluating a Lock, T		•
using semapl	nore, Moi	nitors, Persistence - File Organization: The i-node,	Crash Consisten	cy file security.
Module:5	VIDTI	JAL MACHINES		2 hour
		Ms Taxonomy of VMs.		2 Hour
Module:6		S OF VIRTUALIZATION		4 hour
		Full Virtualization with binary translation, Hardwa	are assisted Oner	
		isted /Para virtualization.	ire assisted, Oper	aning System
, II tauliZuliO	, 00 400.	and a manufaction.		
Module:7	HYPE	RVISOR		5 hour
		virtualization, Server Virtualization, Desktop Virtu	alization. Overvi	
		apshots, OVF, Hot and Cold Cloning Protecting In		
		ainer / Docker.	3	<i>37 0 6 7</i>
Module:8		ENT TRENDS		2 hour
	KECI	ZITI IMELIDIJ		
	1	Total Lecture hours:	30 hours	
		Total Lecture nours:	ou nours	

Text Book(s)

- 1. Silberschatz, Abraham, Greg Gagne, and Peter B. Galvin, "Operating system concepts", 10<sup>th</sup> Edition, Wiley Publishers, 2018.
- 2. Matthew Portnoy, "Virtualization Essentials", John Wiley Sons Inc; 2<sup>nd</sup>Edition Edition, 2016.

#### **Reference Books**

- 1. Thomas Anderson, Michael Dahlin, "Operating Systems: Principles and Practice", 2<sup>nd</sup> Edition, Recursive Books, 2014.
- 2. William Stallings, "Operating Systems: Internals and Design Principles", 8th Edition, 2014.
- 3. Smith, Nair, "Virtual Machines: Versatile Platforms for Systems and Processes", 1st Edition, Morgan Kaufmann Publishers, 2005.

#### Mode of Evaluation: CAT / Assignment / Quiz / FAT / LAB / Seminar

#### **List of Indicative Experiments**

- 1. Study of Basic Linux Commands.
- 2. | Shell Programming (I/O, Decision making, Looping, Multi-level branching).
- 3. Crating child process using fork() system call, Orphan and Zombie process creation.
- 4. Simulation of CPU scheduling algorithms (FCFS, SJF, Priority and Round Robin).
- 5. Simulation of Bankers algorithm to check weather given system is in safe state or not. Also check whether addition resource requested can be granted immediately.
- 6. Parallel Thread management using pthread library. Implement a data parallelism using multi-threading.
- 7. Dynamic memory allocation algorithms first-fit, best-fit, worst-fit algorithms.
- 8. Page Replacement Algorithms FIFO, LRU and Optimal.
- 9. Virtualization Setup: Type-1, Type-2 Hypervisor.
- 10. Implementation of OS / Server Virtualization.

	<b>1</b>									
	Total Laboratory Hours 30 hours									
Mode of assessment: CAT / Assignment / Quiz / FAT / Seminar										
]	Recommended by Board of Studies 13-05-2016									
A	approved by Academic Council	No. 41	Date	17-06-2016						

CSE6006	NOSQL Databases	L	T	P	J	C
		2	0	2	4	4
Pre-requisite	NIL	Syl	lab	us v	ers	ion
						1.0

- 1. Explore the origins of NoSQL databases and the characteristics that distinguish them from traditional relational database management systems.
- 2. Understand the architectures and common features of the main types of NoSQL databases (key-value stores, document databases, column-family stores, graph databases)
- 3. Discuss the criteria that decision makers should consider when choosing between relational and non-relational databases and techniques for selecting the NoSQL database that best addresses specific use cases.

#### **Expected Course Outcome:**

- 1. Explain the detailed architecture, Database properties and storage requirements
- 2.Differentiate and identify right database models for real time applications
- 3. Outline Key value architecture and characteristics
- 4. Design Schema and implement CRUD operations, distributed data operations
- 5. Compare data ware housing schemas and implement various column store internals
- 6.Choose and implement Advanced columnar data model functions for the real time applications
- 7. Develop Application with Graph Data model

#### Module:1 INTRODUCTION TO NOSQL CONCEPTS

4hours

Data base revolutions: First generation, second generation, third generation, Managing Transactions and Data Integrity, ACID and BASE for reliable database transactions, Speeding performance by strategic use of RAM, SSD, and disk, Achieving horizontal scalability with Data base sharding, Brewers CAP theorem.

# Module:2 NOSQL DATA ARCHITECTURE 4 hours PATTERNS

NoSQL Data model: Aggregate Models- Document Data Model- Key-Value Data Model- Columnar Data Model, Graph Based Data Model Graph Data Model, NoSQL system ways to handle big data problems, Moving Queries to data, not data to the query, hash rings to distribute the data on clusters, replication to scale reads, Database distributed queries to Data nodes.

Module:3   KEY VALUE DATA STORES 5 hours			
	Module:3	KEY VALUE DATA STORES	5 hours

From array to key value databases, Essential features of key value Databases, Properties of

keys, Characteristics of Values, Key-Value Database Data Modeling Terms, Key-Value Architecture and implementation Terms, Designing Structured Values, Limitations of Key-Value Databases, Design Patterns for Key-Value Databases, Case Study: Key-Value Databases for Mobile Application Configuration

#### Module:4 DOCUMENT ORIENTED DATABASE

4hours

Document, Collection, Naming, CRUD operation, querying, indexing, Replication, Sharding, Consistency Implementation: Distributed consistency, Eventual Consistency, Capped Collection, Case studies: document oriented database: Mongo DB and/or Cassandra

#### Module:5 COLUMNAR DATA MODEL - I

4 hours

Data warehousing schemas: Comparison of columnar and row-oriented storage, Column-store Architectures: C-Store and Vector-Wise, Column-store internals and, Inserts/updates/deletes, Indexing, Adaptive Indexing and Database Cracking.

#### Module:6 | COLUMNAR DATA MODEL – II

**3hours** 

Advanced techniques: Vectorized Processing, Compression, Write penalty, Operating Directly on Compressed Data Late Materialization Joins, Group-by, Aggregation and Arithmetic Operations, Case Studies

#### Module:7 DATA MODELING WITH GRAPH

4 hours

Comparison of Relational and Graph Modeling, Property Graph Model Graph Analytics: Link analysis algorithm- Web as a graph, Page Rank- Markov chain, page rank computation, Topic specific page rank (Page Ranking Computation techniques: iterative processing, Random walk distribution Querying Graphs: Introduction to Cypher, case study: Building a Graph Database Application- community detection

Module:8 Recent trends 1 hours

Total Lecture hours: 3

30 hours

#### **Reference Books**

- 1. Christopher D.manning, Prabhakar Raghavan, Hinrich Schutze, An introduction to Information Retrieval, Cambridge University Press
- 2. Daniel Abadi, Peter Boncz and Stavros Harizopoulas, The Design and Implementation of Modern Column-Oriented Database Systems, Now Publishers.
- 3. Guy Harrison, Next Generation Database: NoSQL and big data, Apress.

Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar

#### **List of Challenging Experiments (Indicative)**

1. ImporttheHubwaydataintoNeo4jandconfigureNeo4j.Then, answer the following questions using the Cypher Query Language:

5 hours

- a) List top 10 stations with most outbound trips (Show station name and number of trips)
- b) List top10 stations with most in bound trips (Show station name and number of trips)
- c) List top 5 routes with most trips (Show starting station name, ending station name and number of trips) (4) List the hour number(forexample13means1pm-2pm) and number of trips which start

	from the station "B.U.Cent d) List the hour number(for exa which end at the station "B	ample 13 mean	s 1pm-2pm)	and number of trips	
2.	Download a zip code dataset at http import to import the zip code dataset answer the following questions by states that have a city called "BOST Find all the states and cities whose n Each city has several zip code number of zip codes and rank t populations.	o://media.mongoet into Mongo using aggrega FON".  ames include to the ci	DB. After tion pipelin the string "B	importing the data, es: (1) Find all the GOST".	5 hours
3.	Mongo DB can query on spa Create a database that stores roa Each car has a maximum perfor the following: Test Cassandras r	d cars. Cars h	nave a man maximum		5 hours
4.	Consistency models.  Master Data Management using Ne The world of master data is changin are swapping their relational databa data. This switch enables them new insights in existing data, pro	ng. Data archite ases with graph to use a data	ects and app databases t store optin	olication developers o store their master mized to discover	5 hours
5.	data and answer questions about  Shopping Mall case study using ordering items from themal land ordered items.	data relations	ships in rea	d time e many customers	5 hours
	al Laboratory Hours				30 hours
	ode of assessment: <i>Project/Activia</i> commended by Board of	13.05.2016			
	idies	10.00.2010			
Ap	proved by Academic Council	No. 41	Date	17.06.2016	

CSE6014	Programming for Data Science	L	T	P	J	C
		0	0	4	0	2
Pre-requisite	NIL	Syllabus version				sion
		3			1.0	

1. To provide necessary knowledge on how to manipulate data objects, produce graphics, analyse data using common statistical methods and generate reproducible statistical reports with programming in Python and R

#### **Expected Course Outcome:**

- 1. Ability to solve the analytical problems using Python and R
- 2. Develop competency in the Python programming language and a number of data- related Python libraries such as Pandas, Numpy, and Scipy
- 3. Ability to communicate results of analysis effectively using visualizations in Python and R
- 4. Import, export and manipulate data and produce statistical summaries of continuous and categorical data in Python and R
- 5. Ability to perform exploratory data analysis using Python and R

Module:1	Expressions, Operators, matrices, Decision Statements in python	2 hours
Module:2	Control Flow and Functions in python	2 hours
Module:3	Classes, Objects, Packages and Files in python	2 hours
Module:4	Tuple, Lists, Sequences, Dictionaries, Comprehensions	2 hours
Module:5	Numpy Arrays objects, Creating Arrays, basic operations, Indexing, Slicing and iterating, copying arrays, shape manipulation, Identity array, eye function, Universal function	2 hours
Module:6	Linear algebra with Numpy, eigen values and eigen vectors with Numpy	2 hours
Module:7	Aggregation and Joining, Pandas Object: Concatenating and appending data frames, index objects	2 hours
	·	
Module:8	Handling Time series data using pandas Handling missing values using pandas	2 hours
Module:9	Reading and writing the data including JSON data	2 hours
	-	

Module:10	Web scraping using python, Combining and merging	2 hours
Module:11		3 hours
	Data transformations	
	Basic matplotlib plots, common plots used in statistical analysis in python	
Module:12	Common plots used in statistical analysis in python Datatypes in R2. Sequence generation, Vector and subscript, Random2 number generation in R  Data frames and R functions2  Data manipulation and Data Reshaping using plyr, dplyr,2	2 hours
	reshape2	
	Parametric statistics and Non-parametric statistics2	
	Continuous and Discrete Probability distribution using R2	
Module:13	Correlation and covariance, contingency tables2	3 hours
	Overview of Sampling, different sampling techniques2	
	R and data base connectivity2	
Module:14	Web application development with R using Shiny2	2 hours
	Approaches to dealing with missing data in R2	
	Exploratory data analysis with simple visualizations using R 2	
	Feature or Attribute selection using R2	
	Dimensionality Reduction with R2	
	Time series data analysis with R2	
	Total Lecture hours:   30 hours	
	Total Lecture hours: 30 hours	

#### Reference Books

- James Payne, "Beginning Python: Using Python 2.6 and Python 3.1" Wrox, Ist Edition, 2010
- 2. Michael T. Goodrich, Roberto Tamassia, Michael H. Gold wasser, "Data Structures and Algorithms in Python", John Wiley & sons, 2013.
- 3. Ivan Idris, "Python Data Analysis", Packt Publishing Limited, 2014
- 4. Wes McKinney, "Python for Data Analysis Data Wrangling with Pandas, NumPy, and IPython", O'Reilly Media, Ist Edition, 2012
- 5. Michael Heydt, "Learning Pandas Python Data Discovery and Analysis Made Easy", Packt Publishing Limited, 2015.
- 6. Jacqueline Kazil, Katharine Jarmul, "Data Wrangling with Python: Tips and Tools to Make Your Life Easier", O'Reilly Media, Ist Edition, 2016.
- 7.https://docs.scipy.org/doc/numpy-dev/reference/index.html#reference
- 8.http://www.python-course.eu/numpy.php
- 9. Michael J. Crawley, "The R Book", Wiley, 2nd Edition, 2012.
- 10. Robert Kabacoff, "R in Action", Manning Publication, Ist Edition, 2011.
- 11. Torsten Hothorn, Brian S. Everitt, "A Handbook of Statistical Analyses Using R", Chapman and Hall\_CRC, 2nd Edition, 2009.
- 12. Chris Beeley "Web Application Development with R Using Shiny", PactPublishing, 2013. 13. Phil Spector, "Data Manipulation with R", Springer, 2008.

	<ul> <li>14. Prabhanjan N. Tattar, Suresh R", wiley, 2016</li> <li>15. Pawel Cichosz, "Data Minin 16. Bater Makhabel, "Learning I</li> </ul>	g Algorithms: E	xplained U	Using R", wiley, 2014
Mode	e of assessment: Project/Activity			
Reco	mmended by Board of Studies	13.05.2016		
Appr	oved by Academic Council	No. 41	Date	17.05.2016

CSE6016	INFORMATION VISUALIZATION		L	T	P	J	C
			2	0	2	4	4
Pre-requisite	NIL	Syllabus version				ion	
							1.0

- 1. To understand the various types of data, apply and evaluate the principles of data visualization.
- 2. Acquire skills to apply visualization techniques to a problem and its associated dataset.
- 3. To apply structured approach to create effective visualizations.
- 4. To learn how to bring valuable insight from the massive dataset using visualization.
- 5. To learn how to build visualization dashboard to support decision making.
- 6. To create interactive visualization for better insight using various visualization tools.

#### **Expected Course Outcome:**

- 1. Identify the data types and its associated visualization mechanisms.
- 2. Apply the various scalar and vector visualization techniques to create suitable visualization for real life applications.
- 3. Handle and analyse multidimensional data and hierarchical data for visualization.
- 4. Perform multivariate data analysis and visualization.
- 5. Apply the visualization guidelines for effective information visualization.
- 6. Demonstrate the concept of visualization through dashboard creation for various applications.
- 7. Choose appropriate methods for the given real world problems and produce meaningful visualization.

Module:1	Introduction to Data Visualization	4 hours
	f data visualization - Data Abstraction - Task Abstraction - Ana on, Human Visual Perception	lysis: Four Levels
Module:2	Visualization Techniques – I	3 hours
Scalar and p	point techniques – vector visualization techniques – matrix visua	lization
Module:3	Visualization Techniques – II	6 hours
Visualizatio	on Techniques for Trees, Graphs, and Networks, Multidimension	nal data
Module:4	Visual Analysis of data from various domains – I	5 hours
Time-orient	ed data visualization – Spatial data visualization and case studie	es
Module:5	Visual Analysis of data from various domains – II	5 hours
Text data vi	sualization – Multivariate data visualization, and case studies	
Module:6	<b>Designing Effective Visualizations</b>	2 hours
Guidelines	for designing successful visualizations, Data visualization dos a	nd don'ts
Module:7	Dashboard Creation and Visual Story Telling	3 hours

Dashboard Design principles, Effective Dashboard Display Media, Dashboard creation using visualization tools for the use cases: Finance- marketing-insurance-healthcare etc.,

Module:8	Recent Trends		2 hours
	Total Lecture hours:	30 hours	

#### **Reference Books**

- 1. Tamara Munzer, "Visualization Analysis and Design", CRC Press, 2014.
- 2. Stephen Few, "Now You See It", Analytics Press, 2009.
- 3. Stephen Few, "Information Dashboard Design: the effective visual communication of data", Oreilly, 2006.
- 4. Matthew O. Ward, Georges Grinstein, Daniel Keim"Interactive Data Visualization: Foundations, Techniques, and Applications", CRC Press, Second Edition, 2015.
- 5. Dr.Chun-hauh Chen, W.K.Hardle, A. Unwin, "Handbook of Data Visualization", Springer publication, 2008.
- 6. Ben Fry, "Visualizing Data", O'Reilly Media, 2008
- 7. Winston Chang, "R Graphics Cookbook", O'Reilly, 2012.
- 8. http://www.fusioncharts.com/whitepapers/

Mod	de of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar	
List	t of Challenging Experiments (Indicative)	
1.	Association Rule Mining and Clustering using R	3 hours
2.	Visualization on KNN or Naïve Bayes Classification using R	3 hours
3.	Financial analysis using Clustering, Histogram and HeatMap	2 hours
4.	Time-series analysis –stock market	2 hours
5.	Visualization of various massive dataset-Finance-Healthcare- Census -Geospatial	2 hours
6.	Market-Basket Data analysis-visualization	2 hours
7.	Text visualization using web analytics	2 hours
8.	Hadoop and R integration in Table au using Hortonworks	2 hours
9.	Google API with maps	2 hours
10.	VisualizationusingD3.js	2 hours
11.	Visualization using Zeppelin	2 hours

12.	12. Network Visualization using Gephi					
13.	13. Visualization of reconstruction network using Qlickview					
14.	Dash Board Creation using Table	eau			2 hours	
				Total Laboratory Hours	30 hours	
Mo	de of assessment: <i>Project/Acti</i>	vity				
Rec Stu	ommended by Board of dies	13.05.2010	6			
App	proved by Academic Council	No. 41	Date	17.06.2016		

Pre-requisite Nil Syllabus ve  Course Objectives:  1. To provide comprehensive knowledge on developing and applying machine learning algorithms for massive real-world datasets in distributed frameworks.  2. To demonstrate the use of big data analytics tools like Spark and Mahout for mining massive datasets.  3. To impart in depth knowledge on Deep Learning and Extreme Learning concepts.  Expected Course Outcome:  1.Identify right machine learning / mining algorithm for handling massive data  2.Apply classification and regression models with Spark and Mahout  3. Implement clustering models using Spark and Mahout  4. Mine social Network graphs using MapReduce  5. Apply semi supervised learning for clustering and classification  6. Use Extreme Learning Machine for classification and regression.  8. Use big data analytics tools such as Spark, Mahout and H2O in solving problems based on Machine learning  Module:1 MapReduce Based Machine Learning 7 Fr.  K-Means, PLANET, Parallel SVM, Association Rule Mining in MapReduce, Inverted Index Ranking, Expectation Maximization, Bayesian Networks  Module:2 Classification and Regression models with Spark and Mahout  Linear support vector machines - Naive Bayes model- Decision Trees - Least square regred Decision trees for regression.  Module:3 Clustering in Spark and Mahout 4 Heirarchical Clustering in a Euclidean and Non-Euclidean Space - The Algorithm of Bradlayyad, and Reina - A variant of K-means algorithm - Processing Data in BFR Algorithm CU	CSE6017	MINING MASSIVE DA	TA	I	T	PJ	C
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2. To demonstrate the use of big data analytics tools like Spark and Mahout for mining massive datasets. 3. To impart in depth knowledge on Deep Learning and Extreme Learning concepts.  Expected Course Outcome:  1. Identify right machine learning / mining algorithm for handling massive data 2. Apply classification and regression models with Spark and Mahout 3. Implement clustering models using Spark and Mahout 4. Mine social Network graphs using MapReduce 5. Apply semi supervised learning for clustering and classification 6. Use deep learning to solve real-life problem 7. Use Extreme Learning Machine for classification and regression. 8. Use big data analytics tools such as Spark, Mahout and H2O in solving problems based on Machine learning  Module:1 MapReduce Based Machine Learning 7 Kr. Means, PLANET, Parallel SVM, Association Rule Mining in MapReduce, Inverted Index Ranking, Expectation Maximization, Bayesian Networks  Module:2 Classification and Regression models with Spark and Mahout  Linear support vector machines - Naive Bayes model- Decision Trees - Least square regres Decision trees for regression.  Module:3 Clustering in Spark and Mahout 4 Heierarchical Clustering in a Euclidean and Non-Euclidean Space - The Algorithm of Bradfayyad, and Reina - A variant of K-means algorithm - Processing Data in BFR Algorithm CU	*			Tearming	g		
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Introduction to Semi-Supervised Learning, Semi-Supervised Clustering, Transductive Support

Introduction, Deep Neural Networks, Deep Belief Networks, Auto Encoders, Recurrent

3 hours

4 hours

Properties of Graphs

Vector Machines

Networks

Module:6 Deep Learning

Module:5 Semi-Supervised Learning

Mod	lule:7	Extreme Learning			2 hours			
Extre	me Lea	rning Machines (ELM), EL	M auto encoder, E	xtreme Su	ipport Vecto	r Regression		
Mod	lule:8	<b>Recent Trends:</b>				2 hours		
				l.				
			Total Lecture ho	ours: 30	hours			
Text	t Book(	(s)						
		Leskovec, Anand Rajaram	nan, Jeffrey Ullm	an, "Min	ing of Mass	sive Datasets",		
		ord Press,2011.	, ,	,	S	,		
	2. Nick	Pentreath, "Machine Learn	ing with Spark", P	ackt Publ	ishing,			
		ier Chapelle, Bernhard Scho	_		-	d Learning", The		
	MIT Pr	ress, 2006.	-		-			
Refe	erence I	Books						
	1. Ron	Bekkerman, Mikhail Bilenk	to, John Langford	"Scaling 1	Up Machine	Learning: Parallel		
	and							
		uted Approaches", Cambrid	•					
		ny Lin, Chris Dyer, "Data-Iı	ntensive Text Proc	essing wi	th MapRedu	ce", Morgan		
	• •	ol Publishers, 2010.						
		nessy, J.L. and Patterson, D.	A., 2011. Comput	er archite	cture: a quan	titative approach.		
	Elsevie					_		
		ndramani Tiwary "Learning	-		-			
		nen Sun, Kar-Ann Toh,		•		xtreme Learning		
	Macmin	nes 2013: Algorithms and A	pplications, Sprii	iger, 2014	•			
	Mode	of Evaluation: CAT / Assign	ment / Ouiz / EAT	Γ / Droject	/ Saminar			
	WIOUC C	n Evaluation. CA1 / Assign	micht / Quiz / TA	1 / 1 10ject	/ Schillar			
List	of Cha	llenging Experiments (Ind	licative)					
1.		ans implementation in Mapl				2 hours		
2.		iation Rule Mining with Ma				2 hours		
3.		on trees in Spark	Produce			2 hours		
4.		payes classification using S <sub>1</sub>	park			2 hours		
5.		aced text processing with Sp				2 hours		
6.		ring models with Spark	<del></del>			2 hours		
7.		ng a recommendation engin	e with Spark			2 hours		
8.		senting social-network data	-			2 hours		
9.	_	menting Semi-supervised Cl				2 hours		
10.		Learning using H2O	<u>0</u>			2 hours		
11.		tive analysis using H2O too	1			2 hours		
12.		Classification using Mahout				2 hours		
13.		al clustering using Mahout				2 hours		
14.		ng a recommendation engin	e with Sparkling v	vater		2 hours		
15.		Learning using DL4J				2 hours		
	P	0 20000 2 2 10	Т	otal Labo	ratory Hou			
Mod	le of as	sessment:		2002 22400	j iiou			
		ded by Board of Studies	13-05-2016					
		by Academic Council	No. 41	Date	17-06-2010	6		
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CSE6018	Streaming Data Analytics		L	T	P	J	C
			2	0	2	4	4
Pre-requisite	Nil	Syllabus version			sion		
							1.0

1. It introduces theoretical foundations, algorithms, methodologies, and applications of streaming data and also provide practical knowledge for handling and analyzing streaming data.

#### **Expected Course Outcome:**

- 1. Recognize the characteristics of data streams that make it useful to solve real-world problems.
- 2. Identify and apply appropriate algorithms for analyzing the data streams for variety of problems.
- 3. Implement different algorithms for analyzing the data streams
- 4. Identify the metrics and procedures to evaluate a model

#### Module:1 Introduction 2 hours

Characteristics of the data streams, Challenges in mining data streams Requirements and principles for real time processing, Concept drift Incremental learning.

#### Module:2 Data Streams 5 hours

Basic Streaming Methods, Counting the Number of Occurrence of the Elements in a Stream, Counting the Number of Distinct Values in a Stream, Bounds of Random Variables, Poisson Processes, Maintaining Simple Statistics from Data Streams, Sliding Windows, Data Synopsis,

Change Detection: Tracking Drifting Concepts, Monitoring the Learning Process

#### Module:3 Decision Trees 4 hours

Very Fast Decision Tree Algorithm (VFDT), The Base Algorithm, Analysis of the VFDT Algorithm, Extensions to the Basic Algorithm: Processing Continuous Attributes, Functional Tree Leaves, Concept Drift.

### Module:4 Clustering from Data Streams 5 hours

Clustering Examples: Basic Concepts, Partitioning Clustering - The Leader Algorithm, Single Pass k-Means, Micro Clustering, Clustering Variables: A Hierarchical Approach

## Module:5 Frequent Pattern Mining 4 hours

Mining Frequent Item sets from Data Streams- Landmark Windows, Mining Recent Frequent Item sets, Frequent Item sets at Multiple Time Granularities

Sequence Pattern Mining- Reservoir Sampling for Sequential Pattern Mining over data streams

#### 4 hours **Module: 6** | Evaluating Streaming Algorithms Evaluation Issues, Design of Evaluation Experiments, Evaluation Metrics, Error Estimators using a Single Algorithm and a Single Dataset, Comparative Assessment, The 0-1 loss function, Evaluation Methodology in Non-Stationary Environments, The Page-Hinkley Algorithm. Module: 7 4 hours **Complex Event Processing** Introduction to Complex Event Processing, Features of CEP, Need for CEP, CEP Architectural Layers, Scaling CEP, Events, Timing and Causality, Event Patterns, Rules and Constraint, STRAW-EPL, Complex Events and Event Hierarchies Module: 8 2 hours **RECENT TRENDS Total Lecture hours:** 30 hours Text Book(s) **Reference Books** 1. Joao Gama, "Knowledge Discovery from Data Streams", CRC Press, 2010. 2. David Luckham, "The Power of Events: An Introduction to Complex Event Processing in Distributed Enterprise Systems", Addison Wesley, 2002. 3. Charu C. Aggarwal, "Data Streams: Models And Algorithms", Kluwer Academic Publishers, 2007 Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar **List of Challenging Experiments (Indicative)** Exploring one stream processing engine like storm or STREAM 1. etc (2 classes) Implementation of algorithms for example: VFDT, CVFDT(2 2. classes) 3 hours each 3. Implementation of Clustering Implementation of Frequent pattern mining 4. Exploring one CEP engine like ESPER or DROOLS(2 classes) 5. Exercise with continuous queries Logical operations on single 6. stream Exercise with continuous queries Logical operations on multiple 7. Exercise with continuous queries temporal operators on single 8. Exercise with continuous queries temporal operators on multiple 9. streams Exercise with complex continuous queries with logical, relational & temporal operators on multiple streams Total Laboratory Hours 30 hours **Mode of assessment:** Recommended by Board of 13.05.2016 **Studies Approved by Academic Council** No. 41 Date 17.06.2016

CSE6019	Text, Web and Social Media Analytics		L	]	ГГ	' J	C
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Pre-requis	ite Nil		Sy	/ll:	abus	ver	sior
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Course Ob	•	1	,	<del></del>			
	ovide an overview of common text mining and social media data derstand the complexities of processing text and network data fr					rcas	
	able students to solve complex real-world problems for sentiment						dati
syste	1	i anarysis a	iiu	1	COII	111011	aati
Expected (	Course Outcome:						
	oret the terminologies, metaphors and perspectives of social medi						
	a wide range of classification, clustering, estimation and predic	tion algori	thm	S C	n Te	xtua	1
data.	rm social network analysis to identify important social actors	cubarou	3C 21	nd	nat	vork	
	rties in social media sites.	, subgroup	os a	Hu	псс	WOIK	
	y state of the art web mining tools and libraries on realistic data s	sets as a ba	sis f	or	busi	ness	
	ons and applications.				_	_	
	de solutions to the emerging problems with social media such	as behavio	or ar	nal	ytics	and	
	nmendation systems.  n new solutions to opinion extraction, sentiment classification an	d data sum	ımar	riz	ation	nrol	oler
o. Desig	in new solutions to opinion extraction, sentiment elassification and	d data sun	iiiiai		atton	prot	<i></i>
Module:1	Introduction to Text Mining					6 h	
Text Repres N-gram mod	entation- tokenization, stemming, stop words, TF-IDF, Feature Veling.	ector Rep	rese	nta	ation	, NE	R,
Module:2	Mining Textual Data					6 h	our
Text Cluster	ing, Text Classification, Topic Modeling-LDA,HDP						
Module:3	Introduction to Web-Mining					6 h	oui
Inverted ind	ces and Boolean queries. PLSI, Query optimization, page rankin	ıg.		_			
Module:4	Web Usage Web content Mining					7 h	our
Web Crawli	ng-Crawler Algorithms, Implementation Issues, Evaluation, Ses	sion & vis	itor	Α	naly	sis, V	/isi
	n, Analysis of Sequential & Navigational Patterns, Predictions b				•		
Module:5	Introduction to Social Media Network					6 h	oui
ssentials of	Social graphs, Social Networks, Models, Information Diffusion i	n Social M	edia	 a.			
	graphic, 2 certa 1 terri oran, 11 certo, antonamicon 2 income			_			
Module:6	Mining Social Media			_		6 h	ou
sehavioral A	nalytics, Influence and Homophily, Recommendation in Social N	Media					
Module:7	Sentimental Mining					6 H	[ou
entiment cla	ssification feature based opinion mining, comparative sentence	and relatio	nal	mi	ning	, Op	inic
oam.							
Module:8	Recent Threads					) L	ou

**Total Lecture hours: 45 hours** 

#### **Reference Books**

- 1. Bing Liu, "Web Data Mining-Exploring Hyperlinks, Contents, and Usage Data", Springer, Second Edition, 2011.
- 2. Reza Zafarani, Mohammad Ali Abbasi and Huan Liu, "Social Media Mining An Introduction", Cambridge University Press, 2014.
- 3. Bing Liu, "Sentiment Analysis and Opinion Mining", Morgan & Claypool Publishers, 2012.
- 4. Nitin Indurkhya, Fred J Damerau, "Handbook of Natural Language Process", 2<sub>nd</sub> Edition, CRC Press, 2010.
- 5. Matthew A.Russell, "Mining the social web", 2nd edition- O'Reilly Media, 2013.

Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar						
Recommended by Board of Studies	13-05-2016					
Approved by Academic Council	No. 41	Date	17-06-2016			

CSE6020	BIG DATA TECHNOLOGIES	L	T	P	J	C
		2	0	2	4	4
Pre-requisite	NIL	Sy	llab	us '	vers	sion
						1.0

- 1. To have knowledge on accessing, storing and manipulating the huge data from different resources.
- 2. To understand the working environment of Pig and Hive for processing the structured and unstructured data.
- 3. To differentiate the RDBMS and Hive architectures and implement queries to process the data using sqoop.
- 4. To have a knowledge on searching mechanisms using solr.

#### **Expected Course Outcome:**

- 1. Illustrate the usage of data on different Big data ecosystems.
- 2. Demonstrate the Pig architecture and evaluation of pig scripts.
- 3. Describe the Hive architecture and execute SQL queries on sample data sets.
- 4. Understand the process of transferring data between different file systems and to execute operations using sqoop.
- 5. Understand the concepts of indexing and use these concepts in solr search engine.
- 6. Implement and evaluate the data manipulation procedures using pig, hive, sqoop and solr.
- 7. Develop an application using different eco system tools by taking standard sample data set.

#### Module:1 Introduction 4 hours

Big data- Concepts, Needs and Challenges of big data. Types and source of big data. Components of Hadoop Eco System- Data Access and storage, Data Intelligence, Data Integration, Data Serialization, Monitoring, Indexing.

# Module:2 Apache Pig 6 hours

Introduction, Parallel processing using Pig, Pig Architecture, Grunt, Pig Data Model-scalar and complex types. Pig Latin- Input and output, Relational operators, User defined functions. Working with scripts.

### Module:3 Apache Hive Fundamentals 3 hours

Introduction-Hive modules, Data types and file formats, Hive QL-Data Definition and Data Manipulation.

Module:4	Apache Hive Advanced Concepts	4 hours
Hive QL que	ries, Hive QL views- reduce query complexity. Hive scr	ipts. Hive QL Indexes-create,
Show drop. A	Aggregate functions. Bucketing vs Partitioning.	

Module:5	Importing and Handling Relational Data in	3 hours
	Hadoop using Sqoop	

Relational database management in Hadoop: Bi directional data transfer between Hadoop and external database. Import data- Transfer an entire table, import subset data, use different file format. Incremental import new data, incrementally import data, preserving the value Module:6 | Sqoop 4 hours Export transfer data from Hadoop, update the data, update at the same time, export subset of columns. Hadoop ecosystem integration- import data to hive, using partitioned hive tables, replace special delimiters. Module:7 4 hours Solr Introduction. Information retrieval search engine, categories of data, inverted index. Design-field attributes and types. Indexing- indexing tool. Indexing operations using CSV documents. Searching data- parameters, default query. Module:8 2 hours **Recent Trends** Total Lecture hours: 30 hours **Reference Books** 1. Alan Gates, Programming Pig Data flow Scripting with Hadoop, O'Reilly Media, Inc,2011. 2. Jason Rutherglen, Dean Wampler, Edward Caprialo, Programming Hive, O'Reilly Media Inc,2012 3. Kathleen Ting, Jarek Jarcec Cecho, Apache Sqoop Cook book, O'Reilly Media Inc, 2013. 4. Dikshant Shahi, Apache Solr: A Practical approach to enterprise search, A press, 2015. 5. Chuck Lam, Hadoop in Action, Manning Publications, 2010. 6. Andrea Gazzarini, Apache Solr Essentials, PACKT Publications, 2015. Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar **List of Challenging Experiments (Indicative)** 6 hours Implement a program using Pig Latin operators and user defined functions Implement a program using operators and Pig Latin scripts Program using Hive manipulation and data definition languages. Implement a program using Hive queries with partitioning. 7 hours Implement a program using Hive indexes. Implement a program using Hive views Implement a program using Hive external table by accessing the external file created by Pig or any other tool. Program using Hive scripts and aggregate functions 6 hours Implement a program using Hive queries with bucketing and clustering. Implement a program for data transfer between Hadoop and external Data base using sqoop. Program to import data and incremental data in sqoop. 6 hours Program to preserve the value in sqoop Program to export data from Hadoop using sqoop Program to import data to hive and using partitioned hive tables 5. 5 hours

Program for inverted index using solr Program for indexing operations using

**Total Laboratory Hours** | 30 hours

CSV files in solr. Program to search data using solr

Mode of assessment: Project/Activity

Recommended by Board of	13.05.2016		
Studies			
Approved by Academic Council	No. 41	Date	17.06.2016

CSE6021		<b>Domain Specific Predictive Analytics</b>		L	T	P		C
Pre-requisi	ito	NIL		3	0	0 bus		4
rre-requisi	ite	NIL		Sy	ша	vus	vei	<b>sion</b> 1.0
Course Ob	jective	:						
		eoretical foundations, algorithms, methodologies	for analys	sing	da	a i	n va	ıriou
domains suc	ch Reta	II, Finance, Risk and Healthcare.						
Expected C	Course	Outcome:						
		lenges in dealing with data sets in domains such as	finance, ri	sk a	nd			
healthcare.	•							
2. Identify	real-wo	orld applications of machine learning in domains su	ıch as finaı	nce,	risk	an	d	
healthcare.								
-		ly appropriate algorithms for analyzing the data for	r variety of	f pro	ble	ms	in	
finance, ris		leattneare.  r a model for new machine learning tasks based on	reasoned a	arom	ner	nt		
4. Wake en	<u> </u>	u model for new machine fourning tasks based on	reasoned t	11541	1101	10		
Module:1	Reta	il Analytics					7 h	ours
Module:2	Risk	Analytics					5 h	ours
	_	t and Operational Hedging: An Overview,						
Managemer and Bankru		ayesian Framework for Supply Chain Risk Manag	gement, Cr	edit	Sco	orin	g	
and Dankiu	picy 11	Aletion						
Module:3	Fina	ncial Data Analytics					5 h	ours
		nalytics: Framework, techniques, and metrics, News	s events in	npac	t m	ark	et	
sentiment,	Relatin	g news analytics to stock returns						
Module:4							<b>6</b> h	ours
		ncial Time Series Analytics	· 1 //D:		<u> </u>			
		dels, Markov chain models, Time series models with						
_	2 1110					~, -		,
forecasting								
	II. al4	a come Amplities					( L	0177
Module:5	Ţ	h care Analytics	D.,	D.		•	6 h	ours
Module:5  Introduction	on to He	h care Analytics ealthcare Data Analytics, Electronic Health Records fethods in Healthcare, Clinical Decision Support S	•	Pres	erv	ing	6 h	ours

Natural Language Processing and Data Mining for Clinical Text: Core NLP Components, Information Extraction and Named Entity Recognition, Social Media Analytics for Healthcare: Tracking of

7 hours

Infectious Disease Outbreaks, Readmission risk Prediction

**Healthcare Data Analytics** 

Module:6

Module:7 Genomic Data Analytics 7 hours

		y Data, Microarray Data A rvival Prediction from Gen	•		A Analysis for Personalized Medicine.  ome Sequence Analysis		
Mod	lule:8	RECENT TRENDS			2 hours		
				•			
			Fotal Lecture h	ours: 4	15 hours		
Text	Boo	$\mathbf{k}(\mathbf{s})$			<u> </u>		
Refe	renc	e Books					
	1. Chris Chapman, Elea McDonnell Feit "R for Marketing Research and Analytics", Springer, 2015.						
	2. (	Dlivia Parr Rud "Data Minin	g Cookbook: M	odeling I	Data for Marketing, Risk, and		
	C	Customer Relationship Mana	agement", Wiley	, 2001.			
	<ol> <li>Chandan K. Reddy, Charu C. Aggarwal "Healthcare Data Analytics", CRC Press, 2015.</li> <li>Rene Carmona "Statistical Analysis of Financial Data in R", Springer, 2014.</li> <li>James B. Ayers "Handbook Of Supply Chain Management" Auerbach Publications, 200</li> </ol>						
	6. P	anos Kouvelis, Ling xiu Do	ong, Onur Boyab	atli, Ron	ng Li "The Handbook of		
	I	ntegrated Risk Management	in Global Supp	ly Chains	s", Wiley, 2012.		
		Evaluation: CAT / Assignm	ent / Quiz / FAT	' / Project	et / Seminar		
		assessment:	<u> </u>				
Reco Stud		ended by Board of	13.05.2016				
		d by Academic Council	No. 41	Date	17.06.2016		
rr		v	1	1	1		

CSE6022	Soft Computing	L	T	P	J	C
		2	0	2	4	4
Pre-requisite	NIL	Sy	llab	us '	vers	sion
						1.0

The objective of this course is to introduce methods for handling imprecise and uncertain data using Rough sets, Neuro Fuzzy Systems and foster their abilities in designing and implementing optimal solutions for real-world and engineering problems using derivative free optimization techniques.

#### **Expected Course Outcome:**

After successfully completing the course the student should be able to

- 1. Have a general understanding of soft computing methodologies, to deal with imprecise and uncertain data
- 2. Develop computational neural network models for some simple biological systems;
- 3. Develop fuzzy models for engineering systems, particularly for control systems;
- 4. Apply derivative free optimization methods to solve real world problems
- 5. Demonstrate some applications of computational intelligence

Module:1	Introduction to Soft Computing	3 hours
Soft Compu	ting Overview - Uncertainty in data, Hard vs Soft Comp	puting
Module:2	Neural Networks	7 hours
Introduction Convolution	n, RBF Networks, Self-Organizing Map, nal Neural Networks	Boltzmann Machines,
Module:3	Fuzzy Systems	3 hours
•	Fuzzy Relations, and Membership functions, Proper n and Defuzzification.	ties of Membership functions,
Module:4	Fuzzy logic	4 hours
Fuzzy Rule b	ased systems, Fuzzy Decision making, Fuzzy Classificati	on, Fuzzy C-Means Clustering.
Module:5	Rough Sets	3 hours
_	<ul> <li>Definition, Upper and Lower Approximations, Bound in Algorithms. Properties of Rough Sets. Rough K-mean ering</li> </ul>	• •

# Module:6 4 hours **Optimization Techniques** Introduction, Genetic Algorithm, Memetic Algorithms, Particle Swarm Optimization, Ant Colony Optimization, Frog-Leaping. **Module:7** | **Hybrid Systems** 4 hours GA Based Back Propagation Networks, Fuzzy Back Propagation Networks, Evolutionary Ensembles Module:8 2 hours **Recent Trends** Total Lecture hours: 30 hours Reference Books **Reference Books** 1. S.N. Sivanandham and S.N.Deepa, "Principles of Soft Computing", 2nd Edition, Wiley Publications. 2. Andries P. Engelbrecht, "Computational Intelligence: An Introduction", John Wiley & Sons.2007 3. Laurene V. Fausett "Fundamentals of Neural Networks: Architectures, Algorithms And Applications", Pearson, 1993 4. Simon Haykin "Neural Networks and Learning Machines" Prentice Hall, 2008. 5. Timothy Ross, "Fuzzy Logic with Engineering Applications", Third Edition, Wiley, Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar **List of Challenging Experiments (Indicative) Project** # Generally a team project consists of four to six members # Down to earth application and innovative idea should have been attempted # Report in Digital format with all drawings using software package to be submitted. # Assessment on a continuous basis with a min of 3 reviews. The following is the sample project that can be given to students to be implemented in any programming languages. • Develop Fuzzy Decision-Making for Job Assignment Problem • Implement TSP using Optimization Techniques

- Develop a suitable method for Health Care Application using Neuro- Fuzzy systems
- Develop a suitable method for Face Recognition System
- Layout Optimization using Genetic Algorithms
- Fault Diagnosis using rough set theory
- Software safety analysis using rough sets

A Neuro-Fuzzy Approach to Bad Debt Recovery in Healthcare

Total Laboratory Hours | 30 hours

Recommended by Board of	13.05.2016		
Studies			
Approved by Academic Council	41	Date	17.06.2016

Pre-requisite Nil Syllabus ver  Course Objectives:  1. To provide students with the fundamentals and essentials of Cloud Computing.  2. To provide students a sound foundation of the Cloud computing so that they are ablestart using and adopting Cloud Computing services and tools in their real life scenarios.  3. To enable students exploring some important cloud computing driven commercial systems as Google Apps, Microsoft Azure and Amazon Web Services and other business cloud applications.  4. To impart knowledge in applications of cloud computing  Expected Course Outcome:  1. Design, Develop & Demonstrate real-world applications from the Cloud Computing  2. Understand the subtle architectural difference in Public and Private Clouds.  3. Appreciate the requirements of various service paradigms in Cloud Computing.  4. Describe the methods of processing multimedia elements and other information	CSE6023	Cloud Computing Fundamentals		L T P J C
<ol> <li>To provide students with the fundamentals and essentials of Cloud Computing.</li> <li>To provide students a sound foundation of the Cloud computing so that they are ablestart using and adopting Cloud Computing services and tools in their real life scenarios.</li> <li>To enable students exploring some important cloud computing driven commercial systems such as Google Apps, Microsoft Azure and Amazon Web Services and other business cloud applications.</li> <li>To impart knowledge in applications of cloud computing</li> <li>Design, Develop &amp; Demonstrate real-world applications from the Cloud Computing</li> <li>Understand the subtle architectural difference in Public and Private Clouds.</li> <li>Appreciate the requirements of various service paradigms in Cloud Computing.</li> <li>Describe the methods of processing multimedia elements and other information</li> </ol>	Pre-requisite	Nil	Sy	2  0  2  4  4 yllabus versio
<ol> <li>To provide students with the fundamentals and essentials of Cloud Computing.</li> <li>To provide students a sound foundation of the Cloud computing so that they are ablestart using and adopting Cloud Computing services and tools in their real life scenarios.</li> <li>To enable students exploring some important cloud computing driven commercial systems such as Google Apps, Microsoft Azure and Amazon Web Services and other business cloud applications.</li> <li>To impart knowledge in applications of cloud computing</li> <li>Design, Develop &amp; Demonstrate real-world applications from the Cloud Computing</li> <li>Understand the subtle architectural difference in Public and Private Clouds.</li> <li>Appreciate the requirements of various service paradigms in Cloud Computing.</li> <li>Describe the methods of processing multimedia elements and other information</li> </ol>	C. Olivai			1.
<ol> <li>To provide students a sound foundation of the Cloud computing so that they are ablestart using and adopting Cloud Computing services and tools in their real life scenarios.</li> <li>To enable students exploring some important cloud computing driven commercial systems such as Google Apps, Microsoft Azure and Amazon Web Services and other business cloud applications.</li> <li>To impart knowledge in applications of cloud computing</li> <li>Design, Develop &amp; Demonstrate real-world applications from the Cloud Computing</li> <li>Understand the subtle architectural difference in Public and Private Clouds.</li> <li>Appreciate the requirements of various service paradigms in Cloud Computing.</li> <li>Describe the methods of processing multimedia elements and other information</li> </ol>				
start using and adopting Cloud Computing services and tools in their real life scenarios.  3. To enable students exploring some important cloud computing driven commercial systems such as Google Apps, Microsoft Azure and Amazon Web Services and other business cloud applications.  4. To impart knowledge in applications of cloud computing  Expected Course Outcome:  1. Design, Develop & Demonstrate real-world applications from the Cloud Computing  2. Understand the subtle architectural difference in Public and Private Clouds.  3. Appreciate the requirements of various service paradigms in Cloud Computing.  4. Describe the methods of processing multimedia elements and other information	1. To provid	le students with the fundamentals and essentials of Cloud	l Computi	ing.
such as Google Apps, Microsoft Azure and Amazon Web Services and other business cloud applications.  4. To impart knowledge in applications of cloud computing  Expected Course Outcome:  1. Design, Develop & Demonstrate real-world applications from the Cloud Computing  2. Understand the subtle architectural difference in Public and Private Clouds.  3. Appreciate the requirements of various service paradigms in Cloud Computing.  4. Describe the methods of processing multimedia elements and other information	-			-
<ol> <li>Design, Develop &amp; Demonstrate real-world applications from the Cloud Computing</li> <li>Understand the subtle architectural difference in Public and Private Clouds.</li> <li>Appreciate the requirements of various service paradigms in Cloud Computing.</li> <li>Describe the methods of processing multimedia elements and other information</li> </ol>	such as C	Google Apps, Microsoft Azure and Amazon Web Service		=
<ol> <li>Design, Develop &amp; Demonstrate real-world applications from the Cloud Computing</li> <li>Understand the subtle architectural difference in Public and Private Clouds.</li> <li>Appreciate the requirements of various service paradigms in Cloud Computing.</li> <li>Describe the methods of processing multimedia elements and other information</li> </ol>	4. To impar	t knowledge in applications of cloud computing		
<ol> <li>Understand the subtle architectural difference in Public and Private Clouds.</li> <li>Appreciate the requirements of various service paradigms in Cloud Computing.</li> <li>Describe the methods of processing multimedia elements and other information</li> </ol>	Expected Cours	e Outcome:		
<ul><li>3. Appreciate the requirements of various service paradigms in Cloud Computing.</li><li>4. Describe the methods of processing multimedia elements and other information</li></ul>	1. Design, I	Develop & Demonstrate real-world applications from the	Cloud Co	omputing
4. Describe the methods of processing multimedia elements and other information	2. Understa	nd the subtle architectural difference in Public and Privat	e Clouds.	
presentation concepts during multimedia communications.	4. Describe	the methods of processing multimedia elements and other	_	_

Cloud Computing Overview: Characteristics – challenges, benefits, limitations, Evolution of Cloud Computing, Cloud computing architecture, Cloud Reference Model (NIST Architecture)

Service Model, Characteristics, Benefits, Enabling Technologies Case Study: AWS, OpenStack

Service Model, Characteristics, Benefits, Enabling Technologies Case Studies: IBM Bluemix,

Service Model, Characteristics, Benefits, Enabling Technologies Case Study: Salesforce.com,

Shared Resources – Resource Pool – Usage and Administration Portal – Usage Monitor – Resource Management – Cloud Security – Workload Distribution – Dynamic provisioning.

4 hours

4 hours

4 hours

5 hours

3 hours

Infrastructure as a Service

Platform as a Service

Software as a Service

Storage as a service

**Data Analytics as a Service** 

Hadoop as a service, MapReduce on Cloud, Chubby locking Service

**Introduction to Public and Private Clouds** 

CRM, Online Collaboration Services

Module:3

Module:4

Module:5

**Module:6** 

Module:7

GAE, Microsoft Azure

Historical Perspective, Datacenter Components, Design Considerations, Power Calculations, Evolution of Data Centers, Cloud data storage - CloudTM Module:8 2 hours **Recent Trends Total Lecture hours:** 30 hours Text Book(s) Reference Books 1) Kai Hwang, Geoffrey Fox, Jack J. Dongarra, Morgan Kaufmann, "Distributed and Cloud Computing: From Parallel Processing to the Internet of Things," 1st Edition, 2011. Shroff. "Enterprise Cloud Computing: Technology, Architecture, Applications", Cambridge press, 2010. 3) Kris Jamsa, "Cloud Computing", Jones & Barlett Learning, 2013. 4) Rajkumar Buyya, James Broberg, Andrzej Goscinski, "Cloud Computing Principles and Paradigms", John Wiley & Sons, 2011. 5) John Rhoton and Risto Haukiojal, "Cloud Computing Architectured: Solution Design Handbook", Recursive Press, 2013. 6) George Recse, "Cloud Application Architectures: Building Application and Infrastructure in the Cloud", O' Reilly Media, First Edition, 2009. 7) Dinkar Sitaram, Geetha Manjunathan, "Moving to the Cloud: Developing Apps in the new world of Cloud Computing", Syngress, 2012. 8) Samee. U. Khan, Albert. Y. Zomaya, "Handbook on Data Centers", Springer, 2015. Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar **List of Challenging Experiments (Indicative)** 30 Hours 1) Cisco simulator – VLAN design, Routing, Sub netting, Gateway configuration 2) Virtual box based Webserver creation, Images/Snapshots access webpage from 2nd VM on another subnet work 3) EC2 AWS – S3 bucket based static webpages. 4) EC2 AWS – Instance Creation, Migration 5) EC2 AWS – Web application using Beanstalk. 6) AWS – Local balancing and auto scaling. 7) IBM Blue Mix - Mobile Application development 8) DaaS – Deployment of a basic web app and add additional functionality(Java scripts based) 9) PaaS – IOT – Mobile sensor based IOT application hosted via PaaS environment 10) SaaS – Deployment of any SaaS application for a online collaborative 11) Deployment of Open stack or Virtual box from the scratch 12) Automating Open stack deployment using Chef/Puppet configuration for 4 node/ 5 node/ HA clusters 13) Hadoop as a Service 14) Cloud TM

15) Online Collaboration Services (User Defined Applications)

		To	tal Labora	tory 30Hours	
Mod	le of assessment:				
Rec	Recommended by Board of Studies 13-05-2016				
App	roved by Academic Council	No. 41	Date	17-06-2016	

CSE6025		Analytics of Things		L T P J C
				3 0 0 4 4
Pre-requis	ite	Nil		Syllabus version 1.0
Course Ob	iective	s:		1.0
		e technology that enables IoT, application of I	oT, cloud su	pport for IoT and
access data	using n	nobile computing devices. This will serve as for		
systems, In	ternet o	f services leading to Industry 4.0 changes.		
Exmested (	7011100	Outcomo		
Expected (		nologies that enables IoT.		
•		dware and software required to design and build	l IoT	
		ms for interfacing with sensors and actuators an		levices
4. Set up th	e serve	rs to upload IoT data to cloud for further analys	is	
Module:1		oduction to IoT		6 hours
		cteristics of IoT, Difference between IoT and M2		
		f IoT, IoT levels and deployment templates, IoT	enabling tec	chnologies: Wireles
Sensor Netw	orks, R	FID, GPS		
Module:2	IOT	Hardware platforms		9 hours
Overview of		pported Hardware Platforms: Raspberry pi, Ard	uino, Intel G	alileo
Module:3	Com	munication in IOT		5 hours
		, Serial, SPI, I2C, 6LoWPAN, 802.11wifi, 802. Ap – Constrained application protocol, RPL		
Module:4	IOT	Software development		7 hours
Linux, Netv		configurations in Linux, Accessing Hardware	& Device Fil	es interactions,
Python pac	kages: J	SON, XML, HTTPLib, URLLib, SMTPLib, X	MPP, Contik	i OS
Module:5	LoT	Dhysical Carryons & Claud Offshings		6 hours
		Physical Servers & Cloud Offerings	Toud of thin	
		oud Storage Models & Communication APIs, Carlo SQL for data processing, WAMP, Designing	`	•
•		vices for IoT	a KESIIUI V	Web AFI, MQ11,
Ailiazoli v	V CO SCI	vices for for		
Module:6	Data	Analytics for IoT		5 hours
Configurir	ng and ı	using Apache Storm for Real-time Data Analysi	S	
Module:7				5 hours
		e Studies illustrating IoT Design		
Smart Hom	e, Smai	rt Parking, weather reporting and monitoring		
Module:8	Recent	Trends		2 hours
	ACCCIII	A I VIIUD		

Total Lecture hours: 45 hou						
Text Book(s)						
Reference Books						
1. Arshdeep Bahga, Vijay Madisetti, "Internet of Things: A hands-on Approach".						
University Press, 2015.						
2. Adrian McEwen & Hakim Cassimally, "Designing the Internet of Things" Wiley, 2014.						
3. Nik Bessis, Ciprian Dobre "Big Data and Internet of Things: A Roadmap for Smart						
Environments", Springer, 2014.						
4. Maik Schmidt "Arduino: A Quick-Start Guide", The Pragmatic Bookshelf, 2011.						
5. Dirk Slama, Frank Puhlmann, Jim Morrish, Rishi M Bhatnagar "Enterprise IoT:						
Strategies and Best Practices for Connected Products and Services", O'Reilly Media,						
2015.						
6. Honbo Zhou, "The Internet of Things in the Cloud: A Middleware Perspective", CRC						
Press, 2012.						
7. Quinton Anderson "Storm Real-time Processing Cookbook", PACKT Publishers, 2013. 8. Onur Dundar, "Home Automation with Intel Galileo", Packt Publishing, 2015						
Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar  Mode of assessment:						
Recommended by Board of Studies   13-05-2016						
Approved by Academic Council No. 41 Date 17-06-2016						

CSE6041	BLOCKCHAIN TECHNOLOGY	L	L T P J				
		2	0	0	4	3	
Pre-requisite	NIL		,	Syll	abu	s versio	
						1.0	
Course Object	ives:						
1.To understand	the technology behind blockchain						
2. To comprehen	d the issues related to blockchain						
3. To study the r	eal-world applications of blockchain						
<b>Expected Cou</b>	rse Outcome:						
A.C	1						
	ly completing the course the student should be able						
	e requirements of the basic design of blockchain						
•	eed of blockchains to find the solution to the real-world p	problems					
	ne working of blockchain	1 1					
•	e underlying technology of transactions, blocks, proof-of-	-work, and	cons	ensu	lS		
building	unlament nevy years of voing blookshain for annications	atlean tlean	4				
	nplement new ways of using blockchain for applications	otner tnan c	rypt	ocui	reno	.y	
	dimensions and the requiere most forms						
	d implement the various platforms						
6. Categorize ar	Introduction					4 hours	
6. Categorize ar  Module:1		in applicatio	n, th	e blo	ocke		
6. Categorize ar  Module:1	Introduction  epts, evolution, structure, characteristics, a sample blockchar	in applicatio	n, th	e blo	ockc	4 hours	

What is a Blockchain? Public Ledgers, Blocks in a Blockchain, Blockchain as public ledgers, Transactions, Distributed consensus. Building a block: Elements of Cryptography-Cryptographic Hash functions, Merkle Tree, Elements of Game Theory

Module:3 Blockchain Architecture and Use cases 5 hours

Design methodology for Blockchain applications, Blockchain application templates, Blockchain application development, Ethereum, Solidity, Sample use cases from Industries, Business problems

Module:4 Smart contracts 4 hours

Smart contract, structure of a contract, interacting with smart contracts using Geth client and Mist wallet, smart contract examples, smart contract patterns

Module:5 Decentralized applications (Dapp	s) 4 hours
---	------------

Dapps,	impleme	nting Dapps, Ethereum Dapps, case studies relate	ed to Dapps			
Modul		Advanced topics		4 hours		
Byzant	ine fault	tolerance, proof-of-work vs proof-of-stake, Secu	urity and Privacy of I	Blockchain, smart		
contrac	t vulnera	bilities, Scalability of Blockchain				
Modul	le:7	Blockchain for real-world applications		4 hours		
Manufa	acturing a	 and production, supply chain management, logist	 ics and transportation	, Internet of things,		
e-voting	g, healtho	care, product life cycle, knowledge and innovatio	n management, new b	usiness models and		
applica	tions		-			
Modul	le:8	Recent Trends		1 hours		
		Total Lecture hours:		30 hours		
Text B	Book(s)					
1.	Blockel	nain applications: a hands-on approach, Bahga A., M	Madisetti V., VPT, 201	7.		
Refere	nce Boo	ks				
<ol> <li>2.</li> </ol>	Singhal Blockel	ing Blockchain, A Beginner's Guide to Buildin , Gautam Dhameja, Priyansu Sekhar Panda, Apress hain A Practical Guide to Developing Business, La	s, 2018.	•		
3.	Blockel	ra and Paul R. Allen, McGraw Hill, 2018.  hain enabled Applications Vikram Dhillon, David usiness Blockchain: Promise, Practice, and Applications	•			
4.	Williar	n Mougayar, Wiley, 2016. hain Science: Distributed Ledger Technology,		-		
5.	Publish	ning; 3rd edition, 2019.				
Mode	of Evalu	ation: CAT / Assignment / Quiz / FAT / Pro	ject / Seminar			
	Project					
	# General	lly an individual project				
	# Concep	ts studied in XXXX should have been used				
	# Down t	o earth application and innovative idea should have	e been attempted			
	# Report	in Digital format with all drawings using software pa	ackage to be submitted.			
	[Ex. 1. D digital clo	resign of a traffic light system using sequential circles.	cuits OR 2. Design of			
	# Assessment on a continuous basis with a min of 3 reviews.					

Projects may be done with focus on real-world applications.

#### Sample Project Titles:

- Implementation of an Automated and Decentralized Pollution
   Monitoring System with Blockchain
- 2. Blockchain-based Malware Detection in Mobile Devices
- 3. Blockchain-Enabled E-Voting
- 4. Blockchain: A Game Changer for Securing IoT Data
- 5. Blockchain-based money transfer
- 6. Stock Market On Blockchain
- 7. Trade Solar Power with neighbours using Blockchain
- 8. Secure Medical Records using Blockchain
- 9. Using Blockchain technology to improve anti-counterfeit measures in different industries
- 10. Blockchain-based land registry
- 11. Blockchain-based loyalty tokens and coins for customers
- 12. Using Blockchain technology for filling up empty hotel rooms
- 13. Secure Blockchain for the art market
- 14. Blockchain for the insurance sector
- 15. Decentralized fleet tracking system, supply chain and logistics

# Mode of assessment: Project/Activity Recommended by Board of 11-06-2019 Studies Approved by Academic No.55 Date 13-06-2019 Council

	Course code Course Title L T P J							
<b>CSE6042</b>		DEEP LEARNING		2 0 2 0 3				
Pre-requis	ite	Nil		Syllabus version				
				V.X.X				
Course Ob								
		oduce the theoretical foundations, algorithm						
		and deep learning. It will help to design and						
_		d also provide the practical knowledge	handling and a	nalysing real world				
applications	s.							
Expected (								
	_	understanding of the fundamental issues a		•				
		fferentiate the concept of machine learning						
		the concept of CNN and transfer learn	ing techniques,	to apply it in the				
		n problems						
		se RNN for language modelling and time s						
		oder and deep generative models to solve p	problems with hi	gh dimensional data				
		kt, image and speech.	•.•	C 1 11				
	-	implement various machine learning al	gorithms in a i	range of real-world				
app	lications.							
M. 1 1. 1	N.T. 1.*		141	00.1				
		ne Learning Basics	4 hours	CO:1				
		, Maximum likelihood estimation, Building						
		r Perceptron, Back-propagation algorithm	and its variants	Stochastic gradient				
decent, Cur	se of Di	mensionality.						
Module:2	Introd	uction to Deep Learning &	5 hours	CO:2				
Module.2		ectures	Silouis	CO.2				
Machine I			ning Width V	s Denth of Neural				
		on Functions: RELU, LRELU, ERELU,		Jachine Learning Vs. Deep Learning, Representation Learning, Width Vs. Depth of Neura				
	1 10 01 , 0001	on runtums, reset, sites of sites of,	o moup or visous	Training of Neural				
Networks, I	Restricte	d Boltzmann Machines, Auto Encoders.		Training of Neural				
Networks, I	Restricte	d Boltzmann Machines, Auto Encoders.		Training of Neural				
•		d Boltzmann Machines, Auto Encoders.  lutional Neural Networks	5 hours					
Module:3	Convo	lutional Neural Networks		CO:3				
Module:3 Architectur	<b>Convo</b>	·		CO:3				
Module:3 Architectur	<b>Convo</b>	<b>lutional Neural Networks</b> view — Motivation - Layers — Filters — F		CO:3				
Module:3 Architectur Popular CN	Convo	<b>lutional Neural Networks</b> view — Motivation - Layers — Filters — F		CO:3				
Module:3 Architectur Popular CN Module:4	Convo	lutional Neural Networks view — Motivation - Layers — Filters — F tectures: ResNet, AlexNet.	arameter sharin	CO:3 ng — Regularization,				
Module:3 Architectur Popular CN Module:4	Convo	lutional Neural Networks view — Motivation - Layers — Filters — F tectures: ResNet, AlexNet.  fer Learning	arameter sharin	CO:3  ng — Regularization,				
Module:3 Architectur Popular CN Module:4	Convo al Overv NN Archi Transi	lutional Neural Networks view — Motivation - Layers — Filters — F tectures: ResNet, AlexNet.  fer Learning	arameter sharing the sharing t	CO:3  ng – Regularization,  CO:3				
Module:3 Architectur Popular CN Module:4 Transfer lea Module:5	Convo	lutional Neural Networks  view — Motivation - Layers — Filters — Fitectures: ResNet, AlexNet.  Fer Learning echniques, Variants of CNN: DenseNet, Pixonce Modelling — Recurrent and sive Nets	4 hours xelNet.  3 hours	CO:3  CO:3  CO:4				
Module:3 Architectur Popular CN  Module:4 Transfer lea  Module:5 Recurrent	Convo	lutional Neural Networks  view — Motivation - Layers — Filters — Fitectures: ResNet, AlexNet.  fer Learning echniques, Variants of CNN: DenseNet, Pixonce Modelling — Recurrent and sive Nets  Networks, Bidirectional RNNs — Encode	4 hours xelNet.  3 hours der-decoder seq	CO:3  CO:4  Quence to sequence				
Module:3 Architectur Popular CN  Module:4 Transfer lea  Module:5 Recurrent	Convo	lutional Neural Networks  view — Motivation - Layers — Filters — Fitectures: ResNet, AlexNet.  Fer Learning echniques, Variants of CNN: DenseNet, Pixonce Modelling — Recurrent and sive Nets	4 hours xelNet.  3 hours der-decoder seq	CO:3  CO:4  Quence to sequence				
Module:3 Architectur Popular CN Module:4 Transfer lea Module:5 Recurrent architechur	Convo	lutional Neural Networks  view — Motivation - Layers — Filters — Fitectures: ResNet, AlexNet.  Ger Learning echniques, Variants of CNN: DenseNet, Pixonce Modelling — Recurrent and sive Nets Networks, Bidirectional RNNs — Encod T for training RNN, Long Short Term Metal	4 hours xelNet.  3 hours der-decoder segmory Networks.	CO:3  CO:4  Quence to sequence				
Module:3 Architectur Popular CN Module:4 Transfer lea Module:5 Recurrent architechur Module:6	Convo	lutional Neural Networks  view — Motivation - Layers — Filters — Fitectures: ResNet, AlexNet.  fer Learning echniques, Variants of CNN: DenseNet, Pixonce Modelling — Recurrent and sive Nets Networks, Bidirectional RNNs — Encoder T for training RNN, Long Short Term Mer	4 hours xelNet.  3 hours der-decoder segmory Networks.	CO:3  CO:3  CO:4  Quence to sequence				
Module:3 Architectur Popular CN  Module:4 Transfer lea  Module:5 Recurrent architechur  Module:6 Under com	Convo al Overv IN Archi  Transf arning Te  Sequer Recurs Neural res - BPT  Auto H aplete Au	lutional Neural Networks  View — Motivation - Layers — Filters — Fitectures: ResNet, AlexNet.  Fer Learning Echniques, Variants of CNN: DenseNet, Pixonce Modelling — Recurrent and sive Nets Networks, Bidirectional RNNs — Encoder T for training RNN, Long Short Term Menuschen Concoders — Regulraized Autoencoders —	4 hours xelNet.  3 hours der-decoder segmory Networks.	CO:3  CO:4  Quence to sequence				
Module:3 Architectur Popular CN Module:4 Transfer lea Module:5 Recurrent architechur Module:6	Convo al Overv IN Archi  Transf arning Te  Sequer Recurs Neural res - BPT  Auto H aplete Au	lutional Neural Networks  View — Motivation - Layers — Filters — Fitectures: ResNet, AlexNet.  Fer Learning Echniques, Variants of CNN: DenseNet, Pixonce Modelling — Recurrent and sive Nets Networks, Bidirectional RNNs — Encoder T for training RNN, Long Short Term Menuschen Concoders — Regulraized Autoencoders —	4 hours xelNet.  3 hours der-decoder segmory Networks.	CO:3  CO:4  Quence to sequence				
Module:3 Architectur Popular CN  Module:4 Transfer lea  Module:5 Recurrent architechur  Module:6 Under com — Contracti	Convolution and Convolution Archivers Transfarming Teachers Recurs Neural res - BPT Auto Fuplete	lutional Neural Networks  View — Motivation - Layers — Filters — Fitectures: ResNet, AlexNet.  Fer Learning Echniques, Variants of CNN: DenseNet, Pixonce Modelling — Recurrent and sive Nets Networks, Bidirectional RNNs — Encoder T for training RNN, Long Short Term Menuschen Concoders — Regulraized Autoencoders —	4 hours xelNet.  3 hours der-decoder segmory Networks.	CO:3  CO:4  Quence to sequence				

2 hours

**CO:6** 

Module:8 Recent Trends

			Total Lecture ho	urs:	30 hours		
Refe	erence l	Books				II.	
1.	1. Ian Goodfellow, Yoshua Bengio and Aaron Courville, "Deep Learning", MIT Press, 2017.					ess, 2017.	
2.	Josh F	atterson, Adam Gibson "D	eep Learning: A P	ractiti	ioner's Appro	ach", O'R	eilly Media,
	2017						
3.		rto Michelucci "Applied D		Case-	based Appro	ach to Ur	derstanding
		Neural Networks" Apress, 2					
4.		P. Murphy "Machine Learn					
5.		Alpaydin,"Introduction to l	Machine Learning'	', MIT	Press, Prent	ice Hall of	India, Third
		n 2014.					
6.		arlo Zaccone, Md. Rezaul K				ning with	TensorFlow:
	_	re neural networks with Pytho				2017	
7.		io Gulli, Sujit Pal "Deep Le					
8.		ois Chollet "Deep Learning				2017.	
		aluation: CAT / Assignmen	t / Quiz / FAT / Pr	oject ,	Seminar		
	ist of Experiments  Classification with Multilayer Perceptron using Scikit-learn (MNIST Dataset) 3 hours						
1.				at-iea	rn (MINIST L	vataset)	3 hours
2.		-Parameter Tuning in Multi		Theore	a and DaTan	a.la	3 hours
3.		earning Packages Basics: T		1 near	10 and PyTor	cn	2 hours
4. 5.		fication of MNIST Dataset	using CIVIN				2 hours
6.	Parameter Tuning in CNN 2 hours						
7.	Sentiment Analysis using CNN 2 hours Face recognition using CNN 2 hours						
8.			coming of CNN on	obitoe	atumas		2 hours
9.	Object detection using Transfer Learning of CNN architectures 2 hours  Recommendation system using Deep Learning 2 hours				2 hours		
							2 hours
10.		nsionality Reduction using I	Deep learning				2 hours
12.		age Modeling using RNN	.Τ				•
		Series Prediction using RNN	<b>\</b>				2 hours
13.		nent Analysis using LSTM					2 hours
14.	image	generation using GAN		т	otal Laborate	ти Цонто	2 hours 30 hours
Mad	lo of arr	aluation: Project/Activity		1	otal Laborate	лу nours	30 Hours
		ded by Board of Studies	11-06-2019				
		<u> </u>		Doto	12 06 0	010	
App	rovea b	y Academic Council	No. 55	Date	13-06-2	W19	

CSE6043	IMAGE AND VIDEO ANALYTICS	L	Т	P	J	С
		2	0	0	4	3
Pre-requisite	NIL	Syllabi	us versi	on		
		1.0				

- 1. To impart knowledge on the basic principles and concepts in digital image and video processing.
- 2. To explore and demonstrate real time image and video analytics in solving practical problems of commercial and scientific interests.

#### **Expected Course Outcome:**

- 1. Understand the requirements of image processing
- 2. Illustrate the principles and techniques of digital image in applications related to digital imaging system
- 3. Demonstrate the image recognition and motion recognition
- 4. Understand the fundamentals of digital video processing
- 5. Illustrate the motion estimation, segmentation and modeling
- 6. Design and Analysis of video processing in application

# Module:1 Introduction 4 hours

Basic steps of Image processing system – Pixel relationship- Image Transforms-. Image Enhancement- Histogram Processing, Spatial filtering, Frequency Domain filtering

Module:2 Image Segmentation, Compression and Colour Image Processing 5 hours

Image Segmentation – Detection of Discontinuities. - Edge Linking and Boundary Detection. - Thresholding. Region-Based Segmentation. Image Compression – Encoder-Decoder model, Lossy and Lossless compression,
Huffman Coding, Arithmetic Coding, JPEG, JPEG 2000. Colour Image Processing – Colour Models, Color
Transformations Color Image Smoothing and Sharpening, Color Noise Reduction, Color-Based Image

Segmentation.

# Module:3 Feature extraction and Texture Analysis 4 hours

Feature Extraction - Binary object feature, Histogram-based (Statistical) Features, Intensity features, Shape feature extraction, PCA - SIFT – SURF. Texture Analysis - Concepts and classification, statistical, structural and spectral analysis.

# Module:4 Object recognition and Image Retrieval 4 hours

Object Recognition -Patterns and pattern class, Bayes' Parametric classification, Feature Selection and Boosting, Template- Matching. Content Based Image Retrieval - Feature based image retrieval, Object Based Retrieval

4 hours
4

Digital Video, Sampling of video signal, Video Enhancement and Noise Reduction- Rate control and buffering, MPEG, H.264, Inter frame Filtering Techniques, Fundamentals of Motion Estimation and Motion Compensation

#### Module:6 Video Segmentation and Tracking

5 hours

Change Detection, Background modelling, Motion Segmentation, Simultaneous Motion Estimation and Segmentation, Motion Tracking, Multi-target/Multi-camera tracking

#### Module:7 Video Analysis Action Recognition

3 hours

Video Analysis Action Recognition, Video based rendering, Context and scene understanding. Case Study: Surveillance - Advanced Driver Assistance System

Module:8	Recent Trends	1 hours
	Total Lecture hours:	30 hours

#### Text Book(s)

- 1. Rafael C. Gonzalez and Richard E. Woods, "Digital Image Processing", Third Ed., Prentice-Hall, 2008
- 2. A. Murat Tekalp, "Digital Video Processing", Second Edition, Prentice Hall, 2015.

#### **Reference Books**

- 1. Oge Marques, "Practical Image and Video Processing Using MATLAB", Wiley-IEEE Press, 2011
- 2. Yu Jin Zhang, "Image Engineering: Processing, Analysis and Understanding", Tsinghua University Press, 2009.
- 3. Mark Nixon and Alberto S. Aquado, "Feature Extraction & Image Processing for Computer Vision", Third Edition, Academic Press, 2012
- 4. Richard Szeliski, "Computer Vision: Algorithms and Applications", Springer, 2010
- 5. Boguslaw Cyganek,"Object Detection and Recognition in Digital Images: Theory and Practice", Wiley 2013

Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar

Mode of assessment: Project/Activity

Project

# Generally a team project [2 to 4 members]

# Application with innovative idea is expected

# Continuous Assessment based on a minimum of 3 reviews.

Sample projects that can be given to students to be implemented

using MATLAB/OpenCV/Python/Octave/C/Java etc:

- 1. Image enhancement applications
- 2. Object/image recognition applications based on digital image transforms
- 3. Image analysis systems for visual inspection tasks (object recognition)
- 4. Image compression, Image Fusion
- 5. Image Steganography, Watermarking
- 6. Applications of Image Intelligence in: Medicine, Microscopy, Remote sensing, Astronomy, Materials science, Security, Robotics, Optical character recognition, Metallography etc
- 7. Defence Smart Surveillance and Tracking
- 8. ADAS Sign Board Detection, Traffic Monitoring, Fatigue Detection, Navigation, Lane detection
- 9.Image Captioning and Visual Question Answering
- 10.Gesture Recognition

#### Links for image database:

- <a href="http://homepages.inf.ed.ac.uk/rbf/CVonline/Imagedbase.htm">http://homepages.inf.ed.ac.uk/rbf/CVonline/Imagedbase.htm</a>
- <a href="https://www.cs.cmu.edu/~cil/v-images.html">https://www.cs.cmu.edu/~cil/v-images.html</a>
- <a href="http://www.imageprocessingplace.com/root\_files\_V3/image\_databases.htm">http://www.imageprocessingplace.com/root\_files\_V3/image\_databases.htm</a>
- https://gengo.ai/datasets/20-best-image-datasets-for-computer-vision

Recommended by Board of	11-06-201	9	
Studies			
Approved by Academic	No.55	Date	13-06-2019
Council			

Course code	Course Title	L T P J C
CSE6046	Network Science and Applications	3 0 2 0 4
Pre-requisite	Nil	Syllabus version
		V.X.X

Introduces network science to an interdisciplinary audience, from the internet, to social networks, and the genetic networks that determine our biological existence

#### **Expected Course Outcome:**

- 1. Understand the need and importance of network science
- 2. Able to represent a network as a graph and introduce the elementary characteristics of networks
- 3. Ability to construct and characterize networks that are truly random and measure its strength and weakness
- 4. Develop a self-consistent theory of evolving networks to predict the dynamics and the topology of a wide range of real networks
- 5. Design a network to ensure the system is robustness and not vulnerable to any attacks. Measure degree correlations and explore their impact on the network topology
- 6. Define and explore various communities and introduce a series of algorithms for community identification.

#### | Module:1 | Introduction | 3 hours |

Vulnerability Due to Interconnectivity, Networks at the Heart of Complex Systems, Two Forces Helped the Emergence of Network Science, The Characteristics of Network Science, Societal Impact, Scientific Impact.

#### **Module:2** | Networks and Graphs

4 hours

C0.2

CO:1

Degree, Average Degree and Degree Distribution, Adjacency Matrix, Real Networks are Sparse, Weighted Networks, Bipartite Networks, Paths and Distances, Connectedness, Clustering Coefficient Advanced Topic - Global Clustering Coefficient.

#### **Module:3** | Random Networks

4 hours

**CO:3** 

Introduction, The Random Network Model, Number of Links, Degree Distribution, Real Networks are Not Poisson, The Evolution of a Random Network, Real Networks are Supercritical, Small Worlds, Clustering Coefficient

#### **Module:4** | Evolving Networks

4 hours

**CO:4** 

Introduction, The Bianconi-Barabasi Model, Measuring Fitness, Bose-Einstein Condensation, Evolving Networks, Initial Attractiveness

#### **Module:5** | **Degree Correlation**

4 hours

**CO:5** 

Introduction, Assortativity and Disassortativity, Measuring Degree Correlations, Structural Cutoffs, Correlations in Real Networks, Generating Correlated Networks, The Impact of Degree Correlations

#### **Module:6** | Network Robustness

4 hours

**CO:5** 

Introduction, Percolation Theory, Robustness of Scale-free Networks, Attack Tolerance, Cascading Failures, Modeling Cascading Failures, Building Robustness

#### **Module:7** Communities

5 hours

**CO:6** 

Introduction, Basics of Communities, Hierarchical Clustering, Modularity, Overlapping Communities, Testing Communities, Characterizing Communities

Spreading Phenomena – Introduction, Epidemic Modeling, Network Epidemics, Contact Networks, Beyond the Degree Distribution, Immunization, Epidemic Prediction

Mod	lule:8	Recent Trends	2 hours		CO:6
		Total Lecture hours:	30 hours		
	Book(	,			
1.		-Laszlo Barabasi, "Network Science", Cambridg	ge university	press	
Dofo	•	1st Edition, 2017.			
1.	rence I	sley and J. Kleinberg, Networks, Crowds and Marke	ats Cambridge	Univ Proce	2010
2.		Newman, Networks: An Introduction, Oxford Universi		Olliv. I less	, 2010.
3.		ndes and T. Erlebach (Eds.), Network Analysis: Me		oundations.	Springer.
	2005.		8	,	~ [
Mod	e of Ev	aluation: CAT / Assignment / Quiz / FAT / Project	/ Seminar		
List	of Cha	llenging Experiments			
1.		ruct different types of real networks and state the n			3 hours
		Compute Degree, Average Degree and Degr	ee Distribution	on for the	
2		ucted networks.	din = 1in1, 1in4,	Datamaina	3 hours
2.	the ax	representation - adjacency matrices, The corresponderage clustering coefficient of the network, court	iding link lists, nt_the_number	Determine of cycles	3 nours
		e, Clustering Coefficient and Components - degree			
3.	Ď	ite Networks	distribution of	THE HELWOIK	3 hours
	-	der a bipartite network			
	Const	ruct its adjacency matrix. Why is it a block-diagona	l matrix?		
	Const	ruct the adjacency matrix of its two projections - Ca	lculate the ave	rage	
	degree				
4.		der a bipartite network with N1 and N2 nodes in the			3 hours
		is the maximum number of links Lmax the network many links cannot occur compared to a non-bipartite		70 N – N1	
	+ N2?	-	e lietwork of si	ZC IN — INI	
	If N1<	< N2, what can you say about the network density, t	hat is the total	number of	
		over the maximum number of links, Lmax?			
		in expression connecting N1, N2 and the average de	gree for the tw	o sets in	
		partite network, $\langle k1 \rangle$ and $\langle k2 \rangle$ .			
		ute global clustering coefficient.			
5.		ruct random networks – number of links – degree di	istributions – c	lustering	3 hours
6.		cient – maximum and minimum degrees ianconi-Barabási Model – calculate degree dynamic	ce Degree diet	ribution	3 hours
0.		rancom-Barabasi Woder – calculate degree dynamic	es, Degree dist	i ioution,	3 Hours
7.		e correlations for any networks – degree correlation	coefficient		3 hours
8.		ning networks that are robust to attacks and random			3 hours
		ate three networks with 104 nodes, that are assortat			
		and have a power-law degree distribution with degree Value Property & Salvalanda and the state of the state o			
		ne Xalvi-Brunet & Sokolov algorithm to generate the			
		computer, study the robustness of the three networks compare their $P_{\infty}(f)/P_{\infty}(0)$ ratio. Which network is the	-		
		n why?	most rooust:	cuii you	
9.	_	ate a random network with the Erdős-Rényi G(N,	p) model and a	a scale-free	3 hours
	netwo	rk with the configuration model, with $N = 103$ nod	es and average	degree (k)	
		ssume that on each node there is a bucket which can	hold as many	sand grains	
	as the	node degree. Simulate then the following process:			

At each time step add a grain to a randomly chosen node i.						
If the number of grains at node i reaches or exceeds its bucket size, then it becomes						
	unstable and all the grains at the r	node topple to the	buckets of	its adjacent nodes.		
	If this toppling causes any of the a	adjacent nodes' bu	ckets to be	e unstable, subsequent		
	topplings follow on those nodes,					
	sequence of toppings an avalanche					
	turned unstable following an initial	•				
	Repeat (a)-(c) 104 times. Assum					
	grains is lost in the transfer, so the		uckets do	not become saturated		
	with sand. Study the avalanche di	stribution P(s)				
10.	Hierarchical Networks - Calculate	e the degree expor	nent - Com	munities on a Circle	3 hours	
	- Calculate the modularity of the	obtained partition	- Modular	ity Resolution Limit		
	– Modularity maximum					
Total: 30						
Mode of evaluation: Project/Activity						
Recommended by Board of Studies 11-06-2019						
Appı	roved by Academic Council	No. 55	Date	13-06-2019		

Course code	Masters Thesis	]	L	T	P	J	С
CSE6099			)	0	0	0	16
Pre-requisite	As per the academic regulations	Syllabus version			sion		
							1.0

To provide sufficient hands-on learning experience related to the design, development and analysis of suitable product / process so as to enhance the technical skill sets in the chosen field and also to give research orientation.

#### **Expected Course Outcome:**

At the end of the course the student will be able to

- 1. Formulate specific problem statements for ill-defined real life problems with reasonable assumptions and constraints.
- 2. Perform literature search and / or patent search in the area of interest.
- 3. Conduct experiments / Design and Analysis / solution iterations and document the results.
- 4. Perform error analysis / benchmarking / costing
- 5. Synthesis the results and arrive at scientific conclusions / products / solution
- 6. Document the results in the form of technical report / presentation

#### **Contents**

- 1. Capstone Project may be a theoretical analysis, modeling & simulation, experimentation & analysis, prototype design, fabrication of new equipment, correlation and analysis of data, software development, applied research and any other related activities.
- 2. Project can be for two semesters based on the completion of required number of credits as per the academic regulations.
- 3. Should be individual work.
- 4. Carried out inside or outside the university, in any relevant industry or research institution.
- 5. Publications in the peer reviewed journals / International Conferences will be an added advantage

Mode of Evaluation: Periodic reviews, Presentation, Final oral viva, Poster submission

Recommended by Board of Studies	13.05.2016		
Approved by Academic Council	41 <sup>st</sup> AC	Date	17.06.2016

MAT6001	ADVANCED STATISTICAL METHODS	L	TP	J	С
		2	0 2	0	3
Pre-requisite	Nil	Sylla	abus	ver	sion
					2.0

- 1. To provide students with a framework that will help them choose the appropriate descriptive statistics in various data analysis situations.
- 2. To analyze distributions and relationships of real-time data.
- 3. To apply estimation and testing methods to make inference and modeling techniques for decision making using various techniques including multivariate analysis.

#### **Expected Course Outcome:**

- 1. Understand the value of statistics as a discipline and its relevance for Engineering
- 2. Analyze data using appropriate graphical methods and numerical summaries
- 3. Interpret and communicate the outcomes of estimation and hypothesis tests in the context of a problem
- 4. Perform large sample test and small sample testing of Hypothesis as well as calculate confidence interval for a population parameter for real time data.
- 5. describe and verify mathematical considerations for analyzing time series, including concepts of white noise, stationary, auto covariance, autocorrelation; apply various techniques of time series models, including the regression with ARMA models

# Module:1 Basic Statistical Tools for Analysis: 4hours

Summary Statistics, Correlation and Regression, Concept of R<sup>2</sup> and Adjusted R<sup>2</sup> and Partial and Multiple Correlation, Fitting of simple and Multiple Linear regression, Explanation and Assumptions of Regression Diagnostics

# Module:2 Statistical inference : 9 hours

Basic Concepts, Normal distribution-Area properties, Steps in tests of significance —large sample tests—Z tests for Means and Proportions, Small sample tests—t-test for Means, F test for Equality of Variances, Chi-square test for independence of Attributes.

# Module:3 Modelling and Forecasting Methods: 9hours

Introduction: Concept of Linear and Non Liner Forecasting model ,Concepts of Trend, Exponential Smoothing, Linear and Compound Growth model, Fitting of Logistic curve and their Applications, Moving Averages, Forecasting accuracy tests.

**Probability models for time series:** Concepts of AR, ARMA and ARIMA models.

Module:4	Design of Experiments:	6hours
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Analysis of variance – one and two way classifications – Principle of design of experiments, CRD – RBD – LSD, Concepts of  $2^2$  and  $2^3$  factorial experiments

Mod	ule:5	Contemporary issues:				2hours	
Lecti	ure by I	ndustry Experts		<u> </u>			
			Total Lecture hou	irs:   30	hours		
Text	Book(s	<u> </u> s)					
		olied Statistics and Probability	for Engineers, 6ed,	(2016),[	Douglas C. M	Montgomery George	
	Stoffe	ne Series Analysis and Its Appr, David S. Springer publicati		amples (	2017), by S	humway, Robert H.,	
Refe	rence I	Books					
		or Hastie and Robert Tibshirance, and Prediction", Second				<i>C</i>	
		usan Milton and Jesse Arnold cations for Engineering and th					
M - 1	C E	-l4' D'-'4-1 A'	O-i- C-uti		4. E:1 A	Total	
Mode	e of Eva	aluation: Digital Assignments	, Quiz, Continuous A	Assessme	ents, Final A	Assessment Test	
List	of Cha	llenging Experiments (Indic	ative)				
1.	Co	mputing Summary Statistics u	ising real time data				
2	Lot	ting and visualizing data usin	g Tabulation and Gra	nphical F	Representati	ons.	
3		plying simple linear and mult nputing and interpreting the c	-				
4.	Tes	sting of hypothesis for Large s	sample tests for real-t	ime prol	blems.		
5.		sting of hypothesis for Small stred comparison (Pre-test and		and Two	Sample me	ean and	
6.	Tes	sting of hypothesis for Small S	Sample tests for F-tes	st			
7	Tes	sting of hypothesis for Small S	Sample tests for Chi-	square te	est		
8	Ap	plying Time series analysis-T	rends. Growth ,Logis	stic, Exp	onential mo	odels	
9	Applying Time series model AR, ARMA and ARIMA and testing Forecasting accuracy tests.						
10							
11	Per	forming 2 <sup>2</sup> factorial experime	ents with real time A	pplicatio	ons		
12	Per	forming 2 <sup>3</sup> factorial experime	ents with real time A	pplicatio	ons		
	Total Laboratory 24 Hours						
		sessment:				•	
		ded by Board of Studies	11.08.2017	<b>N</b> =4:	24.00.15		
Appl	i uvea l	y Academic Council	No. 46	<b>Date</b>	24.08.17		

SET5001	SCIENCE, ENGINEERING AND TECHNOLOGY PROJECT- I		L	T	P	J	С
							2
Pre-requisite		Syl	llabı	us '	Ve	rsic	n
Anti-requisite							1.0

- To provide opportunity to involve in research related to science / engineering
- To inculcate research culture
- To enhance the rational and innovative thinking capabilities

#### **Expected Course Outcome:**

On completion of this course, the student should be able to:

- 1. Identify problems that have relevance to societal / industrial needs
- 2. Exhibit independent thinking and analysis skills
- 3. Demonstrate the application of relevant science / engineering principles

#### **Modalities / Requirements**

- 1. Individual or group projects can be taken up
- 2. Involve in literature survey in the chosen field
- 3. Use Science/Engineering principles to solve identified issues
- 4. Adopt relevant and well-defined / innovative methodologies to fulfill the specified objective
- 5. Submission of scientific report in a specified format (after plagiarism check)

Student Assessment: Periodical reviews, oral/poster presentation						
Recommended by Board of Studies 17-08-2017						
Approved by Academic Council	No. 47	Date	05-10-2017			

SET5002	SCIENCE, ENGINEERING AND TECHNOLOGY		L	T	P	J	C
	PROJECT- II						
							2
Pre-requisite		Syl	lab	ıs '	Vei	rsic	n
Anti-requisite							1.0

- To provide opportunity to involve in research related to science / engineering
- To inculcate research culture
- To enhance the rational and innovative thinking capabilities

#### **Expected Course Outcome:**

On completion of this course, the student should be able to:

- 1. Identify problems that have relevance to societal / industrial needs
- 2. Exhibit independent thinking and analysis skills
- 3. Demonstrate the application of relevant science / engineering principles

# **Modalities / Requirements**

- 1. Individual or group projects can be taken up
- 2. Involve in literature survey in the chosen field
- 3. Use Science/Engineering principles to solve identified issues
- 4. Adopt relevant and well-defined / innovative methodologies to fulfill the specified objective
- 5. Submission of scientific report in a specified format (after plagiarism check)

	-	•	1 0				
Student Assessment: Periodical reviews, oral/poster presentation							
Recommended by Board of Studies	17-08-2017						
Approved by Academic Council	No. 47	Date	05-10-2017				

ENG5001	Fundamentals	of Communication Skills	L T P J C
			0 0 2 0 1
Pre-requisite	Not cleared EPT (English	Proficiency Test)	Syllabus version
			1.0
<b>Course Object</b>			
		on skills - Listening, Speaking, Re	
		tion in social and academic conte	
		dish language through listening a	nd reading
<b>Expected Cou</b>			
	listening and comprehension sk		
	king skills to express their thou	ghts freely and fluently	
	es for effective reading		
_	atically correct sentences in ger		
	nical writing skills like writing	instructions, transcoding etc.,	0.1
Module:1 Li			8 hours
Understanding			
Listening to Sp			
	pecific Information		4 hours
Exchanging Inf	eaking		4 Hours
	vities, Events and Quantity		
	ading		6 hours
Identifying Info			U HOUIS
Inferring Mean			
Interpreting tex	•		
	riting: Sentence		8hours
Basic Sentence	=	I	0110 011
Connectives	~ 11 00 002 0		
Transformation	of Sentences		
Synthesis of Se	ntences		
	riting: Discourse		4hours
Instructions			
Paragraph			
Transcoding			
		Total Lecture hou	rs: 30 hours
Text Book(s)			•
		Gillie Cunningham. Face2face U	pper
Intermedia	te Student's Book. 2013, Cambi	ridge University Press.	
Reference Boo			
		approach to writing sentences ar	nd Paragraphs
,	lition), 2012, Library of Congre		
		nb, Effective Interpersonal and To	
	· ·	3, John Wiley & Sons, Inc., Hobo	•
	•	arya, New Media Communication	ı Skills for
_	and IT Professionals,2012, IGI	•	D 11 1
		iples and Skills, 2016, 5th Edition,	
5. John Lang Press: USA	an, 1en Steps to Improving Col	lege Reading Skills, 2014, 6th Edi	tion, Iownsend
Tess:USA			

Press:USA

6.	Redston, Chris, Theresa Clements Teacher's Book. 2013, Cambridge			Face2face Upp	er Intermediate
	Authors, book title, year of publication	ation, edition nu	mber, press	, place	
Mo	de of Evaluation: CAT / Assignmer	nt / Quiz / FAT	Project / Se	eminar	
	List of Chall	enging Experin	nents (Indi	cative)	
1.	Familiarizing students to adjective all letters of the English alphabet starts with the first letter of their i	and asking then	n to add an a		2 hours
2.	Making students identify their peeduring presentation and respond u		, Clarity and	d Volume	4 hours
3.	Using Picture as a tool to enhance	learners speaki	ng and writi	ing skills	2 hours
4.	Using Music and Songs as tools to language / Activities through VIT			the target	2 hours
5.	Making students upload their Self	f- introduction v	ideos in Vii	meo.com	4 hours
6.	Brainstorming idiomatic expressi writings and day to day conversat		them use th	ose in to their	4 hours
7.	Making students Narrate events badd flavor to their language / Act	y adding more			4 hours
8	Identifying the root cause of stage to make their presentation better				4 hours
9	Identifying common Spelling & S day to day conversations	Sentence errors	n Letter Wr	riting and other	2 hours
10.	Discussing FAQ's in interviews we better insight in to interviews / Ac				2 hours
	1		Total Lab	oratory Hours	32 hours
	de of evaluation: Online Quizzes, P ni Project	resentation, Rol	e play, Groi	up Discussions,	Assignments,
	commended by Board of Studies	22-07-2017			
	proved by Academic Council	No. 46	Date	24-8-2017	

ENG5002		Professional and Communication	n Skills L T P J C
			0 0 2 0 1
<b>Pre-requisite</b>	E	NG5001	Syllabus version
			1.1
Course Obje			
		to develop effective Language and Comm	unication Skills
_,		ts' Personal and Professional skills	
		nts to create an active digital footprint	
Expected Co			
-		personal communication skills	
		lem solving and negotiation skills	
	_	es and mechanics of writing research repo	rts
		er public speaking and presentation skills	
5. Apply	y the acq	uired skills and excel in a professional en	vironment
Module:1	Person	al Interaction	2hour
		ne's career goals	
Activity: SWC			
Module:2	_	ersonal Interaction	2 hours
Interpersonal (Activity: Role		cation with the team leader and colleagues at	the workplace
Module:3		Interaction	2 hour
		ocial Networking, gender challenges	<b>2</b> 11001
		edIn profile, blogs	
Module:4		é Writing	4 hour
Identifying job	b requirer	nent and key skills	
		ectronic Résumé	
Module:5	Intervi	ew Skills	4 hour
Placement/Job	Interviev	w, Group Discussions	
		w and mock group discussion	
Module:6	_	Writing	4 hour
Language and			
Activity: Write Module:7		Skills: Note making	2hours
Summarizing			2hour
Activity: Abst	ract. Exe	cutive Summary, Synopsis	
Module:8		reting skills	2 hour
Interpret data	_		
Activity: Tran		and graphs	
Module:9		tation Skills	4 hour
Oral Presentat	ion using	Digital Tools	
		ion on the given topic using appropriate non-	verbal cues
Module:10	-	m Solving Skills	4 hour
Problem Solvi	ing & Cor	nflict Resolution	
Activity: Case	Analysis	of a Challenging Scenario	
		Total Lecture hours:	30hour
Text Book(s	)		
_ ~~~ ~~ ~~ ~~ ~~ ()	,		
	gar Nitin	and Mamta Bhatnagar, Communicative I	Inglish For Engineers And

Jon Kirkman and Christopher Turk, Effective Writing: Improving Scientific, Technical and Business Communication, 2015, Routledge   Diana Bairaktarova and Michele Eodice, Creative Ways of Knowing in Engineering, 201   Springer International Publishing   Clifford A Whitcomb & Leslie E Whitcomb, Effective Interpersonal and Team   Communication Skills for Engineers, 2013, John Wiley & Sons, Inc., Hoboken: New Jer   Arun Patil, Henk Eijkman &Ena Bhattacharya, New Media Communication Skills for   Engineers and IT Professionals, 2012, IGI Global, Hershey PA.   Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar   List of Challenging Experiments (Indicative)	Refe	erence Books				
Diana Bairaktarova and Michele Eodice, Creative Ways of Knowing in Engineering, 201 Springer International Publishing Clifford A Whitcomb & Leslie E Whitcomb, Effective Interpersonal and Team Communication Skills for Engineers, 2013, John Wiley & Sons, Inc., Hoboken: New Jer Arun Patil, Henk Eijkman &Ena Bhattacharya, New Media Communication Skills for Engineers and IT Professionals, 2012, IGI Global, Hershey PA.  Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar  List of Challenging Experiments (Indicative)  1. WOT Analysis – Focus specially on describing two strengths and two Weaknesses 2. Role Plays/Mime/Skit Workplace Situations 3. Use of Social Media – Create a LinkedIn Profile and also write a page or two on areas of interest 4. Prepare an Electronic Résumé and upload the same in vimeo 5. Group discussion on latest topics 6 Report Writing – Real-time reports 7 Writing an Abstract, Executive Summary on short scientific or research Articles 8 Transcoding – Interpret the given graph, chart or diagram 9 Oral presentation on the given topic using appropriate non-verbal cues 4 hours 10 Problem Solving Case Analysis of a Challenging Scenario 4 hours  Total Laboratory Hours 32 hours Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignmer Mini Project	1	Jon Kirkman and Christopher Tur	rk, Effective Writii	ıg: Impre	oving Scientific,	Technical and
Diana Bairaktarova and Michele Eodice, Creative Ways of Knowing in Engineering, 201 Springer International Publishing Clifford A Whitcomb & Leslie E Whitcomb, Effective Interpersonal and Team Communication Skills for Engineers, 2013, John Wiley & Sons, Inc., Hoboken: New Jer Arun Patil, Henk Eijkman &Ena Bhattacharya, New Media Communication Skills for Engineers and IT Professionals, 2012, IGI Global, Hershey PA.  Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar  List of Challenging Experiments (Indicative)  1. WOT Analysis – Focus specially on describing two strengths and two Weaknesses 2. Role Plays/Mime/Skit Workplace Situations 3. Use of Social Media – Create a LinkedIn Profile and also write a page or two on areas of interest 4. Prepare an Electronic Résumé and upload the same in vimeo 5. Group discussion on latest topics 6 Report Writing – Real-time reports 7 Writing an Abstract, Executive Summary on short scientific or research Articles 8 Transcoding – Interpret the given graph, chart or diagram 9 Oral presentation on the given topic using appropriate non-verbal cues 4 hours 10 Problem Solving Case Analysis of a Challenging Scenario 4 hours  Total Laboratory Hours 32 hours Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignmer Mini Project						
Springer International Publishing Clifford A Whitcomb & Leslie E Whitcomb, Effective Interpersonal and Team Communication Skills for Engineers, 2013, John Wiley & Sons, Inc., Hoboken: New Jer Arun Patil, Henk Eijkman &Ena Bhattacharya, New Media Communication Skills for Engineers and IT Professionals, 2012, IGI Global, Hershey PA.  Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar  List of Challenging Experiments (Indicative)  1. WOT Analysis – Focus specially on describing two strengths and two Weaknesses 2. Role Plays/Mime/Skit Workplace Situations 3. Use of Social Media – Create a LinkedIn Profile and also write a page or two on areas of interest 4. Prepare an Electronic Résumé and upload the same in vimeo 5. Group discussion on latest topics 6 Report Writing – Real-time reports 7 Writing an Abstract, Executive Summary on short scientific or research Articles 8 Transcoding – Interpret the given graph, chart or diagram 9 Oral presentation on the given topic using appropriate non-verbal cues 4 hours 10 Problem Solving Case Analysis of a Challenging Scenario  Total Laboratory Hours Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignmer Mini Project	2					
Communication Skills for Engineers, 2013, John Wiley & Sons, Inc., Hoboken: New Jer Arun Patil, Henk Eijkman &Ena Bhattacharya, New Media Communication Skills for Engineers and IT Professionals, 2012, IGI Global, Hershey PA.  Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar  List of Challenging Experiments (Indicative)  1. WOT Analysis – Focus specially on describing two strengths and two Weaknesses  2. Role Plays/Mime/Skit Workplace Situations  3. Use of Social Media – Create a LinkedIn Profile and also write a page or two on areas of interest  4. Prepare an Electronic Résumé and upload the same in vimeo  5. Group discussion on latest topics  6 Report Writing – Real-time reports  7 Writing an Abstract, Executive Summary on short scientific or research Articles  8 Transcoding – Interpret the given graph, chart or diagram  9 Oral presentation on the given topic using appropriate non-verbal cues  10 Problem Solving Case Analysis of a Challenging Scenario  4 hours  Total Laboratory Hours  32 hours  Mode of evaluation: Online Quizzes, Presentation, Role play, Group Discussions, Assignmer Mini Project					o o	
Arun Patil, Henk Eijkman &Ena Bhattacharya, New Media Communication Skills for Engineers and IT Professionals, 2012, IGI Global, Hershey PA.  Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar  List of Challenging Experiments (Indicative)  1. WOT Analysis – Focus specially on describing two strengths and two Weaknesses  2. Role Plays/Mime/Skit Workplace Situations 4 hours  3. Use of Social Media – Create a LinkedIn Profile and also write a page or two on areas of interest  4. Prepare an Electronic Résumé and upload the same in vimeo 2 hours  5. Group discussion on latest topics 4 hours  6 Report Writing – Real-time reports 2 hours  7 Writing an Abstract, Executive Summary on short scientific or research 4 hours  8 Transcoding – Interpret the given graph, chart or diagram 2 hours  9 Oral presentation on the given topic using appropriate non-verbal cues 4 hours  10 Problem Solving Case Analysis of a Challenging Scenario 4 hours  Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignmer Mini Project	3	Clifford A Whitcomb & Leslie E	Whitcomb, Effect	ive Interp	personal and Tea	ım
Engineers and IT Professionals, 2012, IGI Global, Hershey PA.   Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar   List of Challenging Experiments (Indicative)						
Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar	4					Skills for
List of Challenging Experiments (Indicative)  1. WOT Analysis – Focus specially on describing two strengths and two Weaknesses  2. Role Plays/Mime/Skit Workplace Situations  3. Use of Social Media – Create a LinkedIn Profile and also write a page or two on areas of interest  4. Prepare an Electronic Résumé and upload the same in vimeo  5. Group discussion on latest topics  6 Report Writing – Real-time reports  7 Writing an Abstract, Executive Summary on short scientific or research Articles  8 Transcoding – Interpret the given graph, chart or diagram  9 Oral presentation on the given topic using appropriate non-verbal cues  10 Problem Solving Case Analysis of a Challenging Scenario  Total Laboratory Hours  32 hours  Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignmen Mini Project						
1. WOT Analysis – Focus specially on describing two strengths and two Weaknesses       2 hours         2. Role Plays/Mime/Skit Workplace Situations       4 hours         3. Use of Social Media – Create a LinkedIn Profile and also write a page or two on areas of interest       2 hours         4. Prepare an Electronic Résumé and upload the same in vimeo       2 hours         5. Group discussion on latest topics       4 hours         6 Report Writing – Real-time reports       2 hours         7 Writing an Abstract, Executive Summary on short scientific or research Articles       4 hours         8 Transcoding – Interpret the given graph, chart or diagram       2 hours         9 Oral presentation on the given topic using appropriate non-verbal cues       4 hours         10 Problem Solving Case Analysis of a Challenging Scenario       4 hours         Total Laboratory Hours       32 hours         Mode of evaluation: Online Quizzes, Presentation, Role play, Group Discussions, Assignment Mini Project		<u> </u>	-	roject / S	eminar	
Weaknesses  2. Role Plays/Mime/Skit Workplace Situations 3. Use of Social Media – Create a LinkedIn Profile and also write a page or two on areas of interest  4. Prepare an Electronic Résumé and upload the same in vimeo  5. Group discussion on latest topics  6 Report Writing – Real-time reports  7 Writing an Abstract, Executive Summary on short scientific or research Articles  8 Transcoding – Interpret the given graph, chart or diagram  9 Oral presentation on the given topic using appropriate non-verbal cues  10 Problem Solving Case Analysis of a Challenging Scenario  Total Laboratory Hours  32 hours  Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignmer Mini Project						
<ol> <li>Role Plays/Mime/Skit Workplace Situations</li> <li>Use of Social Media – Create a LinkedIn Profile and also write a page or two on areas of interest</li> <li>Prepare an Electronic Résumé and upload the same in vimeo</li> <li>Group discussion on latest topics</li> <li>Report Writing – Real-time reports</li> <li>Writing an Abstract, Executive Summary on short scientific or research Articles</li> <li>Transcoding – Interpret the given graph, chart or diagram</li> <li>Problem Solving Case Analysis of a Challenging Scenario</li> <li>Total Laboratory Hours</li> <li>Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignmer Mini Project</li> </ol>	1.		on describing two s	trengths	and two	2 hours
3.       Use of Social Media – Create a LinkedIn Profile and also write a page or two on areas of interest       2 hours         4.       Prepare an Electronic Résumé and upload the same in vimeo       2 hours         5.       Group discussion on latest topics       4 hours         6       Report Writing – Real-time reports       2 hours         7       Writing an Abstract, Executive Summary on short scientific or research Articles       4 hours         8       Transcoding – Interpret the given graph, chart or diagram       2 hours         9       Oral presentation on the given topic using appropriate non-verbal cues       4 hours         10       Problem Solving Case Analysis of a Challenging Scenario       4 hours         Total Laboratory Hours       32 hours         Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignment Mini Project						
on areas of interest  4. Prepare an Electronic Résumé and upload the same in vimeo  5. Group discussion on latest topics  6 Report Writing – Real-time reports  7 Writing an Abstract, Executive Summary on short scientific or research Articles  8 Transcoding – Interpret the given graph, chart or diagram  9 Oral presentation on the given topic using appropriate non-verbal cues  10 Problem Solving – Case Analysis of a Challenging Scenario  Total Laboratory Hours  Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignmen Mini Project	2.	2. Role Plays/Mime/Skit Workplace Situations			4 hours	
4.Prepare an Electronic Résumé and upload the same in vimeo2 hours5.Group discussion on latest topics4 hours6Report Writing – Real-time reports2 hours7Writing an Abstract, Executive Summary on short scientific or research Articles4 hours8Transcoding – Interpret the given graph, chart or diagram 92 hours9Oral presentation on the given topic using appropriate non-verbal cues 4 hours4 hours10Problem Solving Case Analysis of a Challenging Scenario4 hoursTotal Laboratory Hours32 hoursMode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignment Mini Project	3.	3. Use of Social Media – Create a LinkedIn Profile and also write a page or two			te a page or two	2 hours
5. Group discussion on latest topics 4 hours 6 Report Writing – Real-time reports 2 hours 7 Writing an Abstract, Executive Summary on short scientific or research Articles 8 Transcoding – Interpret the given graph, chart or diagram 2 hours 9 Oral presentation on the given topic using appropriate non-verbal cues 4 hours 10 Problem Solving Case Analysis of a Challenging Scenario 4 hours  Total Laboratory Hours 32 hours  Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignment Mini Project		on areas of interest				
6 Report Writing – Real-time reports 2 hours 7 Writing an Abstract, Executive Summary on short scientific or research Articles 8 Transcoding – Interpret the given graph, chart or diagram 9 Oral presentation on the given topic using appropriate non-verbal cues 4 hours 10 Problem Solving Case Analysis of a Challenging Scenario 4 hours  Total Laboratory Hours Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignment Mini Project				in vimeo		
7 Writing an Abstract, Executive Summary on short scientific or research Articles  8 Transcoding – Interpret the given graph, chart or diagram 9 Oral presentation on the given topic using appropriate non-verbal cues 4 hours 10 Problem Solving Case Analysis of a Challenging Scenario 4 hours  Total Laboratory Hours Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignment Mini Project						4 hours
Articles  8  Transcoding – Interpret the given graph, chart or diagram 2 hours  9  Oral presentation on the given topic using appropriate non-verbal cues 4 hours  10  Problem Solving Case Analysis of a Challenging Scenario 4 hours  Total Laboratory Hours 32 hours  Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignment Mini Project	6	Report Writing – Real-time repor		2 hours		
8 Transcoding – Interpret the given graph, chart or diagram 2 hours 9 Oral presentation on the given topic using appropriate non-verbal cues 4 hours 10 Problem Solving Case Analysis of a Challenging Scenario 4 hours  Total Laboratory Hours Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignment Mini Project	7					4 hours
9 Oral presentation on the given topic using appropriate non-verbal cues 4 hours 10 Problem Solving Case Analysis of a Challenging Scenario 4 hours  Total Laboratory Hours 32 hours  Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignment Mini Project		Articles				
10 Problem Solving Case Analysis of a Challenging Scenario 4 hours  Total Laboratory Hours 32 hours  Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignment Mini Project	8	8 Transcoding – Interpret the given graph, chart or diagram				2 hours
Total Laboratory Hours 32 hours  Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignment Mini Project	9	Oral presentation on the given topic using appropriate non-verbal cues				4 hours
Mode of evaluation: : Online Quizzes, Presentation, Role play, Group Discussions, Assignment Mini Project	10				4 hours	
Mini Project	Total Laboratory Hours					
<u> </u>	Mod	le of evaluation: : Online Quizzes,	Presentation, Role	play, Gr	oup Discussions	, Assignments,
Recommended by Board of Studies 22-07-2017				-		-
	Reco	ommended by Board of Studies	22-07-2017			
Approved by Academic Council No. 47 Date 05-10-2017	App	roved by Academic Council	No. 47	Date	05-10-2017	

		FRANCAIS FONCTION	NEL	L T P J
				2 0 0 0
Pre-requisit	e	Nil		Syllabus versi
G 01.1				
Course Obje				
		ents the necessary background to: competence in reading, writing, and speaking	hasia Eranah inal	ludina knowlodao
		(related to profession, emotions, food, workp		-
famil		(related to profession, emotions, rood, workp	race, sports/11000	ies, classiooni and
	•	ciency in French culture oriented view point.		
Z. Aciii	eve proi	eithey in French culture offented view point.		
Expected Co	ourse O	tcome:		
e students wil				
1. Rem	ember tl	e daily life communicative situations via perso	onal pronouns, em	nphatic
		utations, negations, interrogations etc.		
		unicative skill effectively in French language		
		comprehension of the spoken / written language		
		nd demonstrate the comprehension of some pa	irticular new rang	ge of unseen
	en mate	ials. a clear understanding of the French culture thr	ough the languag	re studied
J. Delli	onstrate	recear understanding of the French culture thi	ough the languag	ge studied.
Module:1	Saluer,	Se présenter, Etablir des contacts		3 hou
Les Salutatio	ns, Les	ombres (1-100), Les jours de la semaine, Les	mois de l'année,	Les Pronoms Sujets,
		s, La conjugaison des verbes réguliers, La cor	njugaison des verl	bes irréguliers- avoir
être / aller / v	enir / fa	re etc.		
35 1 1 2	D (		. 1	
Module:2	Présen	1 1 /		3 hou
	person	oondant(e), Demander des nouvelles d'une	;	
	person			
La c	onjugais	on des verbes Pronc	ominaux,	La Négatio
	3 0	Est-ce que ou sans Est-ce que'.	,	$\mathcal{E}$
L´ınterrogati				
-				
Module:3	Situer	ın objet ou un lieu, Poser des questions		4 hou
Module:3 L'article (déf	<b>Situer</b> fini/ inde	un objet ou un lieu, Poser des questions fini), Les prépositions (à/en/au/aux/sur/dans/a		le contracté, Les heur
Module:3 L'article (déf en français,	<b>Situer</b> fini/ inde La Nati	un objet ou un lieu, Poser des questions fini), Les prépositions (à/en/au/aux/sur/dans/a onalité du Pays, L'adjectif (La Couleur, l'ad	jectif possessif, l	le contracté, Les heur l'adjectif démonstrat
Module:3 L'article (défen français, l'adjectif inte	<b>Situer</b> fini/ inde La Nati errogatif	nn objet ou un lieu, Poser des questions fini), Les prépositions (à/en/au/aux/sur/dans/a onalité du Pays, L'adjectif (La Couleur, l'ad (quel/quelles/quelle/quelles), L'accord des adj	jectif possessif, l	le contracté, Les heur l'adjectif démonstrat
Module:3 L'article (défen français, l'adjectif inte	<b>Situer</b> fini/ inde La Natierrogatif	un objet ou un lieu, Poser des questions fini), Les prépositions (à/en/au/aux/sur/dans/a onalité du Pays, L'adjectif (La Couleur, l'ad	jectif possessif, l	le contracté, Les heur l'adjectif démonstrat
Module:3 L'article (défen français, l'adjectif inte	Situer fini/ inde La Nati errogatif ent/ Com	nn objet ou un lieu, Poser des questions fini), Les prépositions (à/en/au/aux/sur/dans/a onalité du Pays, L'adjectif (La Couleur, l'ad (quel/quelles/quelle/quelles), L'accord des adj oien / Où etc.,	jectif possessif, l	le contracté, Les heur l'adjectif démonstrat
Module:3 L'article (défen français, l'adjectif inte	Situer fini/ inde La Nati errogatif nt/ Com	nn objet ou un lieu, Poser des questions fini), Les prépositions (à/en/au/aux/sur/dans/a onalité du Pays, L'adjectif (La Couleur, l'ad (quel/quelles/quelle/quelles), L'accord des adj	jectif possessif, l	le contracté, Les heur l'adjectif démonstrat n, L'interrogation
Module:3 L'article (défen français, l'adjectif inte avec Comme  Module:4	Situer fini/ inde La Nati errogatif ent/ Com Faire of	In objet ou un lieu, Poser des questions fini), Les prépositions (à/en/au/aux/sur/dans/a bnalité du Pays, L'adjectif (La Couleur, l'ad- (quel/quelles/quelle/quelles), L'accord des adjoien / Où etc.,  es achats, Comprendre un texte court,	jectif possessif, l	le contracté, Les heur l'adjectif démonstrat n, L'interrogation
Module:3 L'article (défen français, l'adjectif inte avec Comme  Module:4 La traduction	Situer Faire ( Demai	nn objet ou un lieu, Poser des questions fini), Les prépositions (à/en/au/aux/sur/dans/a palité du Pays, L'adjectif (La Couleur, l'adjectif (La Couleur, l'adjeuel/quelles/quelle/quelles), L'accord des adjoien / Où etc.,  es achats, Comprendre un texte court, der et indiquer le chemin.  (français-anglais / anglais – français)	jectif possessif, l	le contracté, Les heur l'adjectif démonstrat n, L'interrogation 6 hou
Module:3 L'article (défen français, l'adjectif interavec Comme Module:4 La traduction	Situer  fini/ inde La National Com Faire ( Demain simple	In objet ou un lieu, Poser des questions fini), Les prépositions (à/en/au/aux/sur/dans/a palité du Pays, L'adjectif (La Couleur, l'adjectif (Ua Couleu	jectif possessif, l	le contracté, Les heur l'adjectif démonstrat n, L'interrogation
Module:3 L'article (défen français, l'adjectif inteavec Comme Module:4 La traduction Module:5	Situer fini/ inde La Nati errogatif ent/ Com  Faire of Demain simple  Trouve généra	nn objet ou un lieu, Poser des questions fini), Les prépositions (à/en/au/aux/sur/dans/a pnalité du Pays, L'adjectif (La Couleur, l'ad (quel/quelles/quelle/quelles), L'accord des adj pien / Où etc.,  es achats, Comprendre un texte court, der et indiquer le chemin. (français-anglais / anglais –français)  r les questions, Répondre aux questions les en français.	jectif possessif, lectifs avec le non	le contracté, Les heur l'adjectif démonstrat m, L'interrogation 6 hou 5 hou
Module:3 L'article (défen français, l'adjectif interavec Comme Module:4 La traduction Module:5 L'article Part	Situer fini/ inde La Nati errogatif errogatif ont/ Com  Faire o Demai simple  Trouv généra	nn objet ou un lieu, Poser des questions  fini), Les prépositions (à/en/au/aux/sur/dans/a  palité du Pays, L'adjectif (La Couleur, l'adjectif (La Coul	jectif possessif, lectifs avec le non	le contracté, Les heur l'adjectif démonstrat m, L'interrogation 6 hou 5 hou
Module:3 L'article (défen français, l'adjectif interavec Comme Module:4 La traduction Module:5 L'article Part	Situer fini/ inde La Nati errogatif errogatif ont/ Com  Faire o Demai simple  Trouv généra	nn objet ou un lieu, Poser des questions fini), Les prépositions (à/en/au/aux/sur/dans/a pnalité du Pays, L'adjectif (La Couleur, l'ad (quel/quelles/quelle/quelles), L'accord des adj pien / Où etc.,  es achats, Comprendre un texte court, der et indiquer le chemin. (français-anglais / anglais –français)  r les questions, Répondre aux questions les en français.	jectif possessif, lectifs avec le non	le contracté, Les heur l'adjectif démonstrat m, L'interrogation 6 hou 5 hou
Module:3 L'article (défen français, l'adjectif inte avec Comme  Module:4 La traduction  Module:5 L'article Part	Situer  fini/ inde La Natierrogatif ent/ Com  Faire of Demain simple  Trouve généra titif, Met ées au N	nn objet ou un lieu, Poser des questions  fini), Les prépositions (à/en/au/aux/sur/dans/a  palité du Pays, L'adjectif (La Couleur, l'adjectif (La Coul	jectif possessif, lectifs avec le non	le contracté, Les heur l'adjectif démonstrat m, L'interrogation 6 hou 5 hou
Module:3 L'article (défen français, l'adjectif inteavec Comme Module:4 La traduction Module:5 L'article Part phrases donn Module:6 Décrivez:	Situer Fini/ inde La Nati errogatif errogatif ont/ Com  Faire o Demai a simple  Trouv généra titif, Met ées au M	In objet ou un lieu, Poser des questions fini), Les prépositions (à/en/au/aux/sur/dans/a pralité du Pays, L'adjectif (La Couleur, l'adjeuel/quelles/quelle/quelles), L'accord des adjoien / Où etc.,  es achats, Comprendre un texte court, der et indiquer le chemin. (français-anglais / anglais – français)  r les questions, Répondre aux questions es en français.  ez les phrases aux pluriels, Faites une phrase a français ou Féminin, Associez les phrases.	jectif possessif, lectifs avec le non	le contracté, Les heur l'adjectif démonstrat m, L'interrogation  6 hou  5 hou  més, Exprimez les

4 hours

Module:7

Dialogue:

Comment ecrire un dialogue

a) Réserver un billet de train b) Entre deux amis qui se rencontrent au café c) Parmi les membres de la famille Entre le client et le médecin 2 hours Module:8 **Invited Talk: Native speakers Total Lecture hours:** 30 hours Text Book(s) Echo-1, Méthode de français, J. Girardet, J. Pécheur, Publisher CLE International, Paris 2010. Echo-1, Cahier d'exercices, J. Girardet, J. Pécheur, Publisher CLE International, Paris 2010. Reference Books CONNEXIONS 1, Méthode de français, Régine Mérieux, Yves Loiseau, Les Éditions Didier, 2004. 2 CONNEXIONS 1, Le cahier d'exercices, Régine Mérieux, Yves Loiseau, Les Éditions Didier, 2004. ALTER EGO 1, Méthode de français, Annie Berthet, Catherine Hugo, Véronique M. Kizirian, Béatrix Sampsonis, Monique Waendendries, Hachette livre 2006. Mode of Evaluation: CAT / Assignment / Quiz / FAT

13.05.2016

Date

17.6.2016

No 41

Recommended by Board of Studies

Approved by Academic Council

GER5001	Deutsch für Anfänger	L T P J C
		2 0 0 0 2
Pre-requisite	NIL	Syllabus
		version
		1.0

The course gives students the necessary background to:

- 1. Enable students to read and communicate in German in their day to day life
- 2. become industry-ready
- 3. Make them understand the usage of grammar in the German Language.

#### **Expected Course Outcome:**

The students will be able to

- 1. Create the basics of German language in their day to day life.
- 2. Understand the conjugation of different forms of regular/irregular verbs.
- 3. Understand the rule to identify the gender of the Nouns and apply articles appropriately.
- 4. Apply the German language skill in writing corresponding letters, E-Mails etc.
- 5. Create the talent of translating passages from English-German and vice versa and To frame simple dialogues based on given situations.

Module:1 3 hours

Einleitung, Begrüssungsformen, Landeskunde, Alphabet, Personalpronomen, Verb Konjugation, Zahlen (1-100), W-fragen, Aussagesätze, Nomen – Singular und Plural

#### Lernziel:

Elementares Verständnis von Deutsch, Genus- Artikelwörter

Module:2 3 hours

Konjugation der Verben (regelmässig /unregelmässig) die Monate, die Wochentage, Hobbys, Berufe, Jahreszeiten, Artikel, Zahlen (Hundert bis eine Million), Ja-/Nein- Frage, Imperativ mit Sie

#### Lernziel

Sätze schreiben, über Hobbys erzählen, über Berufe sprechen usw.

Module:3 4 hours

Possessivpronomen, Negation, Kasus- AkkusatitvundDativ (bestimmter, unbestimmterArtikel), trennnbare verben, Modalverben, Adjektive, Uhrzeit, Präpositionen, Mahlzeiten, Lebensmittel, Getränke

#### Lernziel:

Sätze mit Modalverben, Verwendung von Artikel, über Länder und Sprachen sprechen, über eine Wohnung beschreiben.

Module:4 6 hours

Übersetzungen : (Deutsch – Englisch / Englisch – Deutsch)

#### Lernziel:

Grammatik – Wortschatz – Übung

Module:5 5 hours

Leseverständnis, Mindmap machen, Korrespondenz-Briefe, Postkarten, E-Mail

#### **Lernziel:**

Wortschat	zbildung und aktiver Sprach	gebrauch			
37.11.6	7				
Module:6					3 hours
Aufsätze:					
Meine Uni	iversität, Das Essen, mein Fr	reund oder meine H	Freund	in, meine Fan	nilie, ein Fest in
Deutschla				,	,
Module:7					4 hours
Dialoge:			l.		
e) Ges	präche mit Familienmitglied	lern, Am Bahnhof,			
f) Ges	präche beim Einkaufen; in	einem Supermarkt	; in ei	ner Buchhand	lung;
,	inem Hotel - an der Rezeption	-			8,
Treffen im	-	,			
Module:8					2 hours
		1 1 1 0	1 1	· · · · · · · · · · · · · · · · · · ·	
	res/Native Speakers / Feinheite chigen Länder	en der deutschen Spr	acne, E	3asisinformatio	n uber die
		<b>Total Lecture ho</b>	urs:	30 hours	
Text Book	(s)		l.		l
	d A1 Deutsch als Fremdspra	che, Hermann Fun	k, Chr	ristina Kuhn, S	Silke Demme :
2012					
Reference		11.6.0.1.5			~
	Deutsch als Fremdsprache	A1, Stefanie Deng	ler, Pa	ul Rusch, Hel	en Schmtiz, Tanja
Sieber,	-	) full mi	<u> </u>	2012	
	e ,Hartmut Aufderstrasse, Ju				
	SprachlehrefürAUsländer, I				
	Aktuell 1, HartmurtAufderstr	asse, Heiko Bock,	Mecht	thildGerdes, J	utta Müller und
	t Müller, 2010				
ww.goe					
	tsdeutsch.de				
	klett-sprachen.de				
ww.deu	tschtraning.org				
Mode of Ev	valuation: CAT / Assignmen	t / Quiz / FAT			
	ded by Board of Studies	13.05.2016			
	by Academic Council	No. 41	Date	17-06-20	016
11	·				

STS500	1	Essentials of Business Etiqu	iettes	L T P J C
·	• .			3 0 0 0 1
Pre-requi	site			Syllabus version 2.0
Course Obj	octives.			۷.۱
		e students' logical thinking skills		
	-	trategies of solving quantitative ability pro	blems	
		verbal ability of the students	Olema	
		ritical thinking and innovative skills		
<b>Expected Co</b>	our <u>se O</u> i	atcome:		
	-	nts to use relevant aptitude and appropriate lar	nguage to express	themselves
2. To co	mmunica	te the message to the target audience clearly		_
			г	0.1
Module:1		s Etiquette: Social and Cultural		9 hours
	-	te and Writing Company Blogs and		
		l Communications and Planning and gress release and meeting notes		
	Wrung	press release and meeting notes		
Value, Manne	ers Custo	ms, Language, Tradition, Building a blog, Dev	veloping brand me	essage FAOs'.
		Open and objective Communication, Two wa		
audience, Idea	ntifying, (	Gathering Information,. Analysis, Determining	g, Selecting plan, F	Progress check,
		te a short, catchy headline, Get to the Point –s	ummarize your su	bject in the first
paragraph., B	ody – Ma	ke it relevant to your audience,		_
Module:2	Study of	kills – Time management skills		3 hours
Wiodule:2	Study Si	kins – Time management skins		3 Hours
Prioritization,	Procrasti	nation, Scheduling, Multitasking, Monitoring,	Working under p	ressure and adhering
to deadlines				
74 1 1 2	- ·			
Module:3		ation skills – Preparing presentation		7 hours
		ganizing materials and Maintaining		
	question	paring visual aids and Dealing with		
	questioi	15		
10 Tips to pi	repare Po	werPoint presentation, Outlining the content.	Passing the Elev	vator Test, Blue sky
		body and conclusion, Use of Font, Use of Co		
		s, Animation to captivate your audience, Desig		
rules, Dealing	; with inte	erruptions, Staying in control of the questions,	Handling difficult	questions
Madulad	04:4	Alilia II N		11 h
Module:4	_	rative Ability -L1 – Number properties		11 hour
		erages and Progressions and ages and Ratios		
	1 el cent			
l l		ages and Natios		
Number of fa	ctors. Fac		on, Tens digit posit	tion, Averages.
		torials, Remainder Theorem, Unit digit position ithmetic Progression, Geometric Progression, 1		

Mo	dule:5	Reasoning Ability-L1 – A	Analytical Reaso	ning	8 hours
		ement(Linear and circular & Cking/grouping, Puzzle test, Se			Blood Relations,
Mo	dule:6	Verbal Ability-L1 – Voca	abulary Building	5	7 hours
	nonyms d	& Antonyms, One word substi	tutes, Word Pairs,	Spellings,	Idioms, Sentence completion,
			Total Lecture h	ours:	45 hours
Ref	erence l	Books			
1.	•	atterson, Joseph Grenny, Ron When Stakes are High. Banga			Crucial Conversations: Tools for orary
2.	Dale Ca	arnegie,(1936) How to Win Fr	iends and Influence	People. N	lew York. Gallery Books
3.	Scott Pe	eck. M(1978) Road Less Trave	elled. New York Ci	ty. M. Sco	ett Peck.
4.	FACE(2	2016) Aptipedia Aptitude Ency	yclopedia. Delhi. V	Viley publi	cations
5.	ETHNU	JS(2013) Aptimithra. Bangalo	re. McGraw-Hill E	ducation P	Pvt. Ltd.
We	bsites:				
1.	www.c	halkstreet.com			
2.	www.s	killsyouneed.com			
3.	www.n	nindtools.com			
4.	www.t	hebalance.com			
5.	www.e	guru.000			
		valuation: FAT, Assignments ts with Term End FAT (Comp		idies, Role	plays,
		ded by Board of Studies	09/06/2017		
		y Academic Council	No. 45 <sup>th</sup> AC	Date	15/06/2017

STS500	2	Preparing for Industry	7	L T P J C
515000		1 Topuling for moustry		3 0 0 0 1
Pre-requi	site			Syllabus version
-				2.0
Course Obj	ectives:			
		op the students' logical thinking skills		
		the strategies of solving quantitative ability	problems	
		n the verbal ability of the students		
4. T	o enhar	nce critical thinking and innovative skills		
Exmented C		taama.		
Expected Co			usa functions or	. d
	_	students to simplify, evaluate, analyze and ons to simulate real situations to be industry		IU
	Apressie	ons to simulate tear situations to be medistry	ready.	
Module:1	Intervi	iew skills – Types of interview and		3 hours
		iques to face remote interviews and		
	Mock 1	Interview		
		ructured interview orientation, Closed quest		
		ective, Questions to ask/not ask during an in		=
		Phone interview preparation, Tips to custon	mize preparation	for personal
interview, Pr	actice r	ounds		
Module:2	Resum	e skills – Resume Template and Use of		2 hours
Wioduic.2		verbs and Types of resume and		2 1100115
		mizing resume		
Structure of		ard resume, Content, color, font, Introduction	n to Power verb	s and Write up,
		sume, Frequent mistakes in customizing r	esume, Layout	- Understanding
different con	npany's	requirement, Digitizing career portfolio		
Module:3	Emotic	onal Intelligence - L1 – Transactional		12 hours
Wiodule:3		sis and Brain storming and		12 Hours
	•	ometric Analysis and Rebus		
	•	s/Problem Solving		
Introduction.		racting, ego states, Life positions, I	ndividual Brai	nstorming, Group
		bladder Technique, Brain writing, Crawfor		
		bursting, Charlette procedure, Round robin		* *
	-	ore than one answer, Unique ways	<i>U</i> ,	,
Module:4	_	itative Ability-L3 – Permutation-		14 hours
		nations and Probability and Geometry		
		ensuration and Trigonometry and		
	_	thms and Functions and Quadratic		
Counting		ons and Set Theory g, Linear Arrangement, Circular Arrangement	gements Condi	tional Probability
_	_	ependent Events, Properties of Polygon, 2I	-	=
-		es, Simple trigonometric functions, Introdu	_	
_		etion to functions, Basic rules of functions, V	_	
-		probabilities of Quadratic Equations, Basic	-	
	-3.255 W	F Squarante Equations, Busic	522500000000000000000000000000000000000	
Module:5		ning ability-L3 – Logical reasoning and		7 hours

		<b>Data Analysis and Interp</b>	pretation		
		Binary logic, Sequential or on-Advanced, Interpretation			etic, Data Sufficiency, Data
Mo	dule:6	Verbal Ability-L3 – Con Logic	nprehension and		7 hours
		nprehension, Para Jumbles & Inference, (c) Strengthe			
			Total Lecture hou	urs:	45 hours
Ref	ference l	Books			
1.		el Farra and JIST Editors(20 ctive Resume in Just One I			r Letter Book: Write and Use Jist Works
2.	Daniel	Flage Ph.D(2003) The Art n. Pearson			
3.		Allen( 2002) Getting Thing enguin Books.	s done: The Art of	Stress -I	Free productivity. New York
4.	FACE(	2016) Aptipedia Aptitude I	Encyclopedia.Delhi.	Wiley	oublications
5.	ETHN	US(2013) Aptimithra. Bang	galore. McGraw-Hil	1 Educat	tion Pvt. Ltd.
We	ebsites:				
1.	www.c	halkstreet.com			
2.	www.s	killsyouneed.com			
3.	www.n	nindtools.com			
4.	www.t	<u>hebalance.com</u>			
5.	www.e	<u>guru.000</u>			
		valuation: FAT, Assignments with Torm End FAT (C			Role plays,
		nts with Term End FAT (C ded by Board of Studies	omputer Based Test 09/06/2017	l)	
		y Academic Council	.1	Date	15/06/2017