				e of Engineering, Sa		
				2021-22		
			Course	Information		
Progra	mme		M.Tech. (CS and	IT)		
Class, Semester		First Year M. Tech	., Sem I			
Course	Code		5IT501			
Course			Advanced Algorith			
Desired	l Requ	isites:	Computer Algorith	ims		
		G 1		T	(3.5.1.)	
		Scheme 3 Hrs/week	T1	Examination Sch	ESE	Total
Lecture Tutoria		5 ms/week	20	T2 20	60	Total 100
Practic		_	20	20	00	100
Interac		_		Credit	s: 3	
		ı				
			Cours	e Objectives		
1		ercise the Grap				
2			oath computing techn			
3	То сс	mpare the algor	rithms based on perf	ormance and comple	xities	
			0.4 (00)	41 D1 4 T	T 1	
A + + 1	nd of			with Bloom's Taxon	omy Level	
CO1			atudents will be able algorithms with real	<u> </u>		Apply
CO2		<u> </u>	<u> </u>	stance based scenario	<u> </u>	Analyze
CO3				em using graph algor		Create
		,				
Modu	le		Mod	ule Contents		Hours
					ion of Graphs, BFS and	
I					s Growing a Minimum	7
			Algorithms of Kruska		1 41 '.1 GGGD '	
II		0	U		ord Algorithm, SSSP in erence Constraints and	6
111			roofs of Shortest-pat		rence Constraints and	0
					Multiplication, Floyd-	
III					Graphs Flow Networks,	7
					ush-relable algorithms	
					Dynamic Multithreading	
IV				* ·	ultithreaded merge sort	6
- '					es, Symmetric positive-	
		etinite matrices P-Completene	and least-squares ap	*	ND completeness of 1	
V			•	MP-complete proble	NP-completeness and	7
					The traveling-salesman	,
VI					inear programming, The	
	1 -	ubset-sum probl				6
	TC1	TT ~		xt Books		
1					vest, "Introduction to Al	gorithms",
2	1 mirc	Lamon me MI	T Press Cambridge,	London, England		
	<u> </u>					
			Re	eferences		
1	Horro	owitz, Sahni Ra			nputer Science, W. H. Fr	eeman and
1		any Press, New	-		<u> </u>	
			Uga	eful Links		
1 2	To be	declared durin	g the course on the C			

		CO-	PO Mapping			
		Programn	ne Outcomes	(PO)		
	PO1	PO2	PO3	PO4	PO5	PO6
CO1	2					
CO2		3		2	1	
CO3	3				2	1

Assessment Plan based on Bloom's Taxonomy Level					
Bloom's Taxonomy Level	T1	T2	ESE	Total	
Remember	To be used minimum				
Understand	To be used	To be used	To be used	To be used	
	minimum	minimum	minimum	minimum	
Apply	10	10	20	40	
Analyze	5	5	15	25	
Evaluate	5	5	15	25	
Create			10	10	
Total	20	20	60	100	

				lege of Engineering	O- O		
			1	AY 2021-22	пзишеј		
Progra	amme		M.Tech. (CS and	rse Information			
Class,		 ster	First Year M. Te				
Course			5IT502	cii., 5ciii i			
Course	e Nam	ıe	Unix Internal				
Desire	d Req	uisites:	Operating System	m			
Te	aching	g Scheme		Examination	Scheme (Marks)		
Lectur		3 Hrs/week	T1	T2	ESE	Tota	ıl
Tutori		-	20	20	60	100	
Practio		-					
Intera	ction	-		Cre	edits: 3		
			Cor	urse Objectives			
1	To in	nterpret design		ilosophy of the U	nix/Linux OS.		
2			chitecture of Uni				
3	To u		of Linux/Unix.				
				O) with Bloom's T	axonomy Level		
			students will be at		OC	A	1
CO1			cture of Unix/Linux	hy of the Unix/Linu	IX US		ply llyze
CO3		pare various IP		. 05			ılyze
						1	
Modu				odule Contents			Hours
I	G O	eneral Overvion perating System	m Services, Assum	- History, System ption About Hardwa		Perspective,	7
II	A S K	ystem Adminis ernel Data Stru	UNIX OS, Introdu tration Architectur acture, System Adm	ction to system core of UNIX OS, Int			6
III	In su	odes, structure		, directories, conve a new file, allocat			7
IV	P	_	nd transitions, layo	out of system mem	• /	f a Process,	7
V	Pi in pi	rocess, Process	n, signals, procest rograms, the user i Scheduling, system	ss termination, av d of a process, the n call for time, cloc	shell, system Boot		6
VI	T M	ypes of IPCs, l	age Queue, Sema	and IPS (Inter proc phore, MPI, Oper			6
1	Mary	rice I Doch "T	The Design of University	Text Books	DHI 100/		
2				Operating System", pplications", TMG		7.	
Refere							
1	Dece	mber, 2010		Unix IPC", Brian			
2	Three	ads", Pearson,	2nd Edition, Dece				ency and
3	Eric	Raymond, "Ar	t of UNIX Program	ming", Pearson, 1s	t edition, October,	2003	

	1
	Haaful I inka
	Useful Links
	https://nptel.ac.in/courses/106/102/106102132/
1	(Intro to Unix System Calls Part 1/2, Kernel Data Structures, Process structure, Context
1	Switching, Fork, Context-Switch, Process Control Block, Locking, File System Implementation,
	File System Operation)
2	https://onlinecourses.nptel.ac.in/noc19_cs50
2	(Processes, Scheduling in Linux, IPC, thread)
3	https://github.com/suvratapte/Maurice-Bach-Notes
4	https://github.com/mit-pdos/xv6-public
5	https://www.geeksforgeeks.org/introduction-to-unix-system/
6	http://www.di.uevora.pt/~lmr/syscalls.html

		CO-	PO Mapping			
		Programn	ne Outcomes	(PO)		
	PO1	PO2	PO3	PO4	PO5	PO6
CO1	2	2				
CO2			1			
CO3	3			2		

Assessment Plan based on Bloom's Taxonomy Level					
Bloom's Taxonomy Level	T1	T2	ESE	Total	
Remember	To be used minimum				
Understand	To be used minimum				
Apply	10	10	20	40	
Analyze	5	5	15	25	
Evaluate	5	5	15	25	
Create			10	10	
Total	20	20	60	100	

		,	Walchand Coll	ege of Engineerii	ıg, Sangli		
	_		(Government A	ided Autonomous	0.		
				AY 2021-22			
				rse Information			
Progra			M.Tech. (CS a				
Class, Semester		FirstYear M. T	Sech., Sem I				
Course			5IT560	. 1.1			
Course			Research Metho	odology			
Desire	a Keq	uisites:					
Tes	achino	Scheme		Evamination	Scheme (Marks)		
Lectur		2 Hrs/week	LA1	LA2	Lab ESE	Total	
Tutoria		-	30	30	40	100	
Practio		-					
Interac	ction	1 Hrs/Week		Cr	edits: 2		
				irse Objectives			
1				vith scientific meth			
2				d hypothesis for d			
3	To ev			for data and result			
A t the	and af			O) with Bloom's T	laxonomy Level		
CO1			e students will be rea for dissertation			Apply	
CO ₂			collection metho			Analyze	
CO ₂		ulate the resear		743		Design	
	1 01111		- Promounen				
Modu	le		Mo	dule Contents		Hours	
	Iı	ntroduction					
I					ch, Types of Research		
					rch, Various stages o	of	
			ence collection				
			lem and Design		the Problem, Techniqu		
II		_	· ·	•		/1	
		Involved in Defining a Problem, Fundamentals of Research Design, Need for Research Design, Different Research Designs					
		ata Collection			<u>. </u>		
III	C	ollection of I	Primary Data, (Observation Meth	od, Interview Method	1, 4	
				·	thods of Data Collection	n	
		U	Analysis of Da				
17.7					Statistics in Research		
IV					ship, Simple Regression		
		ssociation of		n and Regression	n , Partial Correlation	1,	
		omputers and					
V		-		process Data An	alysis and Visualization	n 4	
,		-		•	s, Plagiarism Checker		
		echnical writi			, 5		
		aper Writing	0	report, Types	of Technical repor	t,	
VI		issertation/the				. 5	
	W	riting. Present	ation technique	es, Patents and other	er IPRs, Tools for repor	t	

writing.

List of Experiments:

- Compare difference between research methodology and research method
- 2. Compare and contrast between basic research and applied research in brief
- 3. Perform the literature survey using following tool:
- 4. Literature Survey Using Web of Science
- 5. Literature Survey Using Scopus
- 6. Design a model for a engineering research
- 7. Compare between model and process in engineering research
- 8. Perform data analysis using modern engineering tools
- 9. Apply the following characteristics of quality research to engineering problem:
 - a) Identifying the problem
 - b) Reviewing literature
 - c) Setting objectives and hypothesis
 - d) Choosing the study of design
 - e) Deciding on the sample design
 - f) Collecting data
 - g) Processing and analyzing data
 - h) Writing the report

	i) Disseminating the findings
	Text Books
1	Kothari C. R, "Research Methodology", 2nd Edition, New Age International, 1990
2	Chopra Deepak and Sondhi Neena, "Research Methodology: Concepts and cases", 2nd
2	Edition, Vikas Publishing House, New Delhi, 2015
	References
1	Melville Stuart and Goddard Wayne, "Research Methodology: An Introduction For Science &
1	Engineering Students", 1st Edition, Kenwyn Juta & Co. Ltd.,1996
2	G. Ramamurthy, "Research Methodology", 2nd Edition, Dream Tech Press, New Delhi, 2015
	Useful Links
1	https://onlinecourses.swayam2.ac.in/ntr21_ed23/preview - Academic Research & Report
1	Writing
2	https://www.scopus.com/search/form.uri?display=basic#basic
3	https://onlinecourses.nptel.ac.in/noc21 ge12/preview - Qualitative Research Methods
3	And Research Writing
4	https://onlinecourses.nptel.ac.in/noc21_hs44/preview - Effective Writing
5	https://webofscienceacademy.clarivate.com/learn
(https://onlinecourses.swayam2.ac.in/ntr21 ed23/preview - Academic Research & Report
6	Writing
7	https://nptel.ac.in/courses/121/106/121106007/

		CO-]	PO Mapping			
Programme Outcomes (PO)						
	PO1	PO2	PO3	PO4	PO5	PO6
CO1	2		1			
CO2				2		1
CO3		3				

Assessment

Assessment Plan based on Bloom's Taxonomy level					
Bloom's Taxonomy Level	LA1	LA2	Lab ESE	Total	
Remember	To be used minimum				
Understand	To be used minimum				
Apply	10	10	10	30	
Analyze	10	10	10	30	
Evaluate	5	5	10	20	
Create	5	5	10	20	
Total	30	30	40	100	

		Walchand College	of Engineering	, Sangli	
		(Government Aide	d Autonomous Ir	istitute)	
		AY	2021-22		
		Course	Information		
Programme		M.Tech. (CS and I	(T)		
Class, Semes	ster	First Year M. Tech.	, Sem I		
Course Code	e	5IT551			
Course Nam	ie	Activity based Lab for Advanced Algorithms			
Desired Req	uisites:	Data Structures, Computer Algorithms			
Teachin	g Scheme	Examination Scheme (Marks)			
Lecture	-	LA1	LA2	Lab ESE	Total
Tutorial	-	30	30	40	100
Practical	2 Hrs/Week				
Interaction	-	Credits: 1			

	Course Objectives					
1	To demonstrate the concepts of Graph Algorithms.					
2	To implement shortest path computing techniques.					
3	To compare the algorithms based on complexities.					
	Course Outcomes (CO) with Bloom's Taxonomy Level					
At the	end of the course, the students will be able to,					
CO1	Demonstrate graph related algorithms with real world problems	Apply				
CO2	Implement the shortest path for a given distance based scenario	Apply				
CO3	Design approximation algorithms in graph	Create				

List of Experiments / Lab Activities

List of Experiments:

Activities are to be carried out individually.

Each student will perform the activity based on course on following areas.

- 1. Implement the Elementary Graph Algorithms and MST
- 2. Demonstrate the Single Source Shortest Path Algorithms
- 3. Implement the Multithreaded Algorithms and Matrix Operations
- 4. Study NP-Completeness and Polynomial-time verification
- 5. Demonstrate the Approximation Algorithms in graph theory

Student should perform the activities on the basis of the real-time applications in the subjects and submit the work with code, PPT, PDF, Text report document & reference material or on online GitHub. Students should maintain activity log book containing weekly progress

online	e GitHub. Students should maintain activity log book containing weekly progress			
	Text Books			
1	Thomas H. Cormen, Charles E. Leiserson and Ronald L. Rivest, "Introduction to			
1	Algorithms", Third Edition the MIT Press Cambridge, London, England.			
	References			
1	Horrowitz, Sahni Rajasekaran, "Computer Algorithms", Computer Science, W. H.			
1	Freeman and company Press, New york			
Useful Links				
1	https://nptel.ac.in/courses/106/101/106101060/			

CO-PO Mapping								
	Programme Outcomes (PO)							
	PO1	PO2	PO3	PO4	PO5	PO6		
CO1	3		1					
CO2		2						
CO3			3		1			

There are three components of lab assessment, LA1, LA2 and Lab ESE.

IMP: Lab ESE is a separate head of passing. LA1, LA2 together is treated as In-Semester Evaluation.

Assessment	Based on	Conducted by	Typical Schedule (for 26-week Sem)	Marks
LA1	Lab activities,	Lab Course	During Week 1 to Week 6	30
LAI	attendance, journal	Faculty	Marks Submission at the end of Week 6	30
LA2	Lab activities,	Lab Course	During Week 7 to Week 12	30
LAZ	attendance, journal	Faculty	Marks Submission at the end of Week 12	30
Lab ESE	Lab activities,	Lab Course	During Week 15 to Week 18	40
LauESE	attendance, journal	Faculty	Marks Submission at the end of Week 18	40

Assessment Plan based on Bloom's Taxonomy level						
Bloom's Taxonomy Level	LA1	LA2	Lab ESE	Total		
Remember	To be used minimum					
Understand	To be used minimum					
Apply	10	10	10	30		
Analyze	10	10	10	30		
Evaluate	5	5	10	20		
Create	5	5	10	20		
Total	30	30	40	100		

Walchand College of Engineering, Sangli				
	(Government Aided Autonomous Institute)			
	AY 2021-22			
	Course Information			
Programme	M.Tech. (CS and IT)			
Class, Semester First Year M. Tech., Sem-1				
Course Code	5IT552			
Course Name Activity based Lab for Unix Internal				
Desired Requisites:	Operating System, (C/python) Programming language			

Teaching Scheme		Examination Scheme (Marks)				
Lecture	-	LA1 LA2 Lab ESE Total				
Tutorial	-	30	30	40	100	
Practical	2 Hrs/Week					
Interaction	-	Credits: 1				

	Course Objectives					
1	To use various system call of Unix/Linux					
2	To elaborate the various inter process communications					
3	To impart the inter process communications for solving the real world prob	lems				
	Course Outcomes (CO) with Bloom's Taxonomy Level					
At the	end of the course, the students will be able to,					
CO1	Illustrate the difference between thread and process	Apply				
CO2	Identify different system calls for Linux/Unix programming	Analyze				
CO3	Implement various inter process communications available in operating					
	system					

List of Experiments / Lab Activities

List of Experiments:

List of Experiments:

- 1. Processing Environment: fork, vfork, wait, waitpid,exec (all variations exec), and exit
- 2. IPC: Interrupts and Signals: signal(any three type of signal), alarm, kill, signal
- 3. File system Internals: Stat, fstat, ustat/lock/flock.
- 4. Threading concept: In c language (P thread) clone, threads of java
- 5. IPC: Semaphore: semaphore. h-semget, semctl, semop

https://github.com/suvratapte/Maurice-Bach-Notes

- 6. IPC: Message Queue: msgget, msgsnd, msgrcv
- 7. IPC: Shared memory: shmget, shmat, shmdt
- 8. IPC: Sockets: socket system calls in C/socket programming of Java/python.
- 9. IPC: Pipe/FIFO
- 10. Scripting writing in Linux and python

Student should perform the activities on the basis of the real-time applications in the subjects and submit the work with code, PPT, PDF, Text report document & reference material or on online GitHub. Students should maintain activity log book containing weekly progress

	Text Books				
1	Maurice J. Bach, "The Design of Unix Operating System", PHI, 1994.				
2	Sumitabha Das, "Unix Concepts and Applications", TMGH, 4 th Edition, 2017.				
	References				
1	Beej Jorgensen, "Beej's Guide to Unix IPC", Brian -Beej Jorgensen Hall, Version				
1	1.1.2, December, 2010				
2	Kay Robbins, Steve Robbins, "UNIX Systems Programming: Communication,				
2	Concurrency and Threads", Pearson, 2nd Edition, December, 2015				
3	Eric Raymond, "Art of UNIX Programming", Pearson, 1st edition, October, 2003				
	Useful Links				
1	https://users.cs.cf.ac.uk/Dave.Marshall/C/				

3	https://github.com/mit-pdos/xv6-public
4	https://www.geeksforgeeks.org/introduction-to-unix-system/
5.	https://github.com/beejjorgensen/bgipc
6.	http://www.di.uevora.pt/~lmr/syscalls.html

CO-PO Mapping								
	Programme Outcomes (PO)							
	PO1	PO2	PO3	PO4	PO5	PO6		
CO1	3		3					
CO2		2				1		
CO3	1			2				

There are three components of lab assessment, LA1, LA2 and Lab ESE.

IMP: Lab ESE is a separate head of passing. LA1, LA2 together is treated as In-Semester Evaluation.

Assessment	Based on	Conducted by	Typical Schedule (for 26-week Sem)	Marks			
LA1	Lab activities, attendance, journal	Lab Course Faculty	During Week 1 to Week 6 Marks Submission at the end of Week 6	30			
LA2	Lab activities, attendance, journal	Lab Course Faculty	During Week 7 to Week 12 Marks Submission at the end of Week 12	30			
Lab ESE	Lab activities, attendance, journal	Lab Course Faculty	During Week 15 to Week 18 Marks Submission at the end of Week 18	40			

Assessment Plan based on Bloom's Taxonomy Level							
Bloom's Taxonomy Level	LA1	LA2	Lab ESE	Total			
Remember	To be used minimum						
Understand	To be used minimum						
Apply	10	10	10	30			
Analyze	10	10	10	30			
Evaluate	5	5	10	20			
Create	5	5	10	20			
Total	30	30	40	100			

	Walchand College of Engineering, Sangli						
	(Government Aided Autonomous Institute)						
			A	Y 2020-21			
				se Information			
Progra	amme		M.Tech. (CS and	IT)			
Class,	Semes	ster	FirstYear M. Tech	., Sem I			
Cours	e Code	2	5IT553				
Cours	e Nam	e	Presentation and T	echnical Report Wr	iting		
Desire	d Req	uisites:					
		g Scheme		1	cheme (Marks)		
Lectur		-	LA1	LA2	Lab ESE	Total	
Tutori		-	30	30 30 40 100			
Practi		-					
Intera	ction	1 Hr/week		Cred	lits: 1		
				rse Objectives			
1				g technical content	t preparation and p	resentation	
2			ort writing tools				
3	To a	nalyze various	relevant practices	of presentation ar	nd report/paper wri	ting	
) with Bloom's Tax	konomy Level		
At the			students will be abl				
CO1		<u> </u>		gures in presentation		Analyze	
CO2	Com	pare suitable t	ools towards pract	ticing write-up and	d presentation	Evaluate	
CO3	CO3 Create effective report and presentations of the technical work Create						
			List of Exper	iments / Lab Activ	ities		

List of Experiments:

PART - A Technical Report Writing

- 1. Experiment 1: Writing technical reports using proper Tense and grammar.
- 2. Experiment 2: Study of various types of technical Reports
 - a. Project report
 - b. Types of Paper(Journal/Conference)
 - c. Indexing(Sci/Scopus)
 - d. Intellectual Property Rights (IPR)
 - e. Selection of paper type for possible publication.
- 3. Experiment 3: Study of technical report Structure I
 - a. Preamble
 - b. Abstract
 - c. Literature review/survey
 - d. Problem statement
 - e. Objectives
- 4. Experiment 3: Study of technical report Structure II
 - a. Methodologies
 - b. Results
 - c. Discussions
 - d. Conclusion
 - e. Acknowledgements
- 5. Experiment 4: Use of Bibliographies/references and proper citations in reports.
- 6. Experiment 5: Use of Citations, referring style and method of using citations.
- 7. Experiment 6: Study of Plagiarism
 - a. Checking plagiarism
 - b. Minimizing plagiarism
- 8. Experiments 7: Learn the Publication Ethics and Peer Review and Mentoring

PART – B Presentation

- 9. PPT's and Animations
- 10. Presentation structure, Number of slides and Time management
- 11. Presentation styles
- 12. Figures and Tables for data representations

Part -C Tools and Practices

- 13. Open Office, Latex, Beamer, Flash, GNU Plot etc.
- 14. End Note; Mendeley, Grammarly, Ginger, 1 Checker, Turnitin etc.

	Text Books						
1	Kothari C. R, "Research Methodology", 2 nd Edition, New Age International, 1990						
2	Chopra Deepak and Sondhi Neena, "Research Methodology: Concepts and cases", 2 nd Edition, Vikas Publishing House, New Delhi, 2015						
	References						
1	Melville Stuart and Goddard Wayne, "Research Methodology: An Introduction For Science & Engineering Students", 1 st Edition, Kenwyn Juta & Co. Ltd.,1996						
2	G. Ramamurthy, "Research Methodology", 2 nd Edition, Dream Tech Press, New Delhi, 2015						
	Useful Links						
1	https://onlinecourses.swayam2.ac.in/ntr21_ed23/preview Academic Research & Report Writing						
2	https://onlinecourses.swayam2.ac.in/cec21_ge18/preview Academic Writing						
3	https://onlinecourses.nptel.ac.in/noc21_ge12/preview Qualitative Research Methods And Research Writing						
4	https://onlinecourses.nptel.ac.in/noc21_hs44/preview Effective Writing						
5	Experiment 8 https://webofscienceacademy.clarivate.com/learn						

CO-PO Mapping							
Programme Outcomes (PO)							
	PO1	PO2	PO3	PO4	PO5	PO6	

CO1	2		1			
CO2	1			3		
CO3		2			3	

There are three components of lab assessment, LA1, LA2 and Lab ESE.

IMP: Lab ESE is a separate head of passing. LA1, LA2 together is treated as In-Semester Evaluation.

Assessment	Based on	Conducted by	Typical Schedule (for 26-week Sem)	Marks
LA1	Lab activities,	Lab Course	During Week 1 to Week 6	30
LAI	attendance, journal	Faculty	Marks Submission at the end of Week 6	30
1.42	Lab activities,	Lab Course	During Week 7 to Week 12	30
LA2	attendance, journal	Faculty	Marks Submission at the end of Week 12	30
Lob ECE	Lab activities,	Lab Course	During Week 15 to Week 18	40
Lab ESE	attendance, journal	Faculty	Marks Submission at the end of Week 18	40

Assessment Plan based on Bloom's Taxonomy level						
Bloom's Taxonomy Level	LA1	LA2	Lab ESE	Total		
Remember	To be used minimum					
Understand	To be used minimum					
Apply	10	10	10	30		
Analyze	10	10	10	30		
Evaluate	5	5	10	20		
Create	5	5	10	20		
Total	30	30	40	100		

	Walchand College of Engineering, Sangli (Government Aided Autonomous Institute)							
	AY 2021-22							
Course Information								
Progra	Programme M.Tech. (CS and IT)							
Class,			First Year M. Tech	., Sem I				
Cours			5IT511					
Cours					rtualization Technique	es		
Desire	ed Rec	quisites:	Computer Networl	KS				
T	1.	C 1		E ' ' C	1 (M. 1.)			
Lectui		g Scheme 3 Hrs/week	T1	Examination So	ESE	Total		
Tutori		3 ms/week	20	T2 20	60	100 100		
Practi		-	20	20	00	100		
Intera		-		Credi	te• 3			
Intera	CHOII	-		Cicui	113. 3			
			Cou	rse Objectives				
1	To e	laborate fundam	nentals of virtualizati					
2				del in cloud computi	ng			
3			ificance of virtualization					
) with Bloom's Tax	onomy Level			
At the	end o		students will be abl					
CO1	Use	service model o	of cloud computing			Apply		
CO2				oy the services on cl	oud infrastructure	Apply		
CO3	Anal	lyze service mo	dels for data center a	applications		Analyze		
Modu				le Contents		Hours		
			Cloud Computing		. 11			
I				, Cloud Reference M		7		
	SAAS, Cloud Deployment Model: Public Cloud, Private Cloud and Hybrid Cloud, Cloud Platforms in Industry							
		irtualization	ationiis iii iiidusti y					
II			are-Meta Server	Virtualization Des	ktop Virtualization,	6		
11			ualization, Storage V		Ktop virtualization,			
		Network Functi						
				Content Delivery Ne	etworks, Resilience			
III				tions: Cloud Firewal	· · · · · · · · · · · · · · · · · · ·	6		
	E	Balancers, Intrus	sion Detection System	ms				
		Virtual Private						
IV				te Subnets, Security	Groups, Network	7		
			List, Network Addre	ss Translation.				
* * *		Cloud Manager		and D / M				
V				outing, Data Manage	ment in Cloud	7		
			ource Management i	III Cloud				
VI			oud Computing	ls, Cloud Simulator, 1	Research trend in			
V 1		•	g, Fog Computing	is, Cloud Simulator,	Research trend in	6		
	1	a Compann	5, 1 05 Companing			1		
			Л	Text Books				
1	Rajk	tumar Buyya, C			"Mastering cloud o	computing", Mc		
1	Grav	w Hill Education	n, 3rd Edition, 2011					
2					ud Computing: Conce	epts, Technology		
	& Architecture", Pearson, 1st Edition, 2010							
	·	1		References	1.0			
1					oud Computing: Conc	epts, Technology		
			arson Prentice Hall,					
2				g: A practical approd	ach for learning and i	mpiementation'',		
	Pear	son, 2nd Edition	11, 2012					

	Useful Links
1	Module: I, II, IV, V, VI https://nptel.ac.in/content/syllabus_pdf/106105167.pdf
2	https://aws.amazon.com/

CO-PO Mapping							
Programme Outcomes (PO)							
	PO1	PO2	PO3	PO4	PO5	PO6	
CO1	1						
CO2			2		2		
CO3		3		1			

Assessment Plan based on Bloom's Taxonomy Level							
Bloom's Taxonomy Level	T1	T2	ESE	Total			
Remember	To be used minimum						
Understand	To be used minimum						
Apply	10	10	20	40			
Analyze	5	5	15	25			
Evaluate	5	5	15	25			
Create			10	10			
Total	20	20	60	100			

			llege of Engineerin					
	(Government Aided Autonomous Institute)							
AY 2021-22 Course Information								
Programme M.Tech. (CS and IT)								
	Class, Semester First Year M. Tech., Sem I							
Course C		5lT512	cii., Seiii i					
Course N			octive 1. Puby & C	Go Programming Language				
	Requisites:	C & CPP Progra		Jo Frogramming Language				
Desireur	requisites.	C & CIT I logia	unning					
Teacl	hing Scheme		Examination	Scheme (Marks)				
Lecture	3 Hrs/week	T1	T1 T2 ESE					
Tutorial	-	20	20	60	Total 100			
Practical	-		<u> </u>					
Interaction			Cı	redits: 3				
	-							
		Co	ourse Objectives					
1 To	o use various parad		Go Programming L	Language				
			ndling and error har					
3 To	o demonstrate the f	eatures of Go lang	guage for process sy	ynchronization				
			CO) with Bloom's T					
	d of the course, the	students will be a	ible to,	•				
CO1 In	nplement object or	iented programmi	ng concepts using F	Ruby	Apply			
CO2 D	emonstrate the con	cept of File handl	ling using Ruby and	l Go language	Apply			
CO3 A	nalyze the Synchro	nization problem	using Go Language	e	Analyze			
Module			odule Contents		Hours			
I	Introduction to Ruby Programming Brief history of Ruby, Installing & running Ruby, Command Line Arguments, Numbers, Text & Strings, Arrays & Hashes, Symbols, Expressions (True, False, Nil) Classes, Modules & Objects: Objects, Classes, Variables							
II	Flow Control & Statements and Properties Conditionals, Loops, Error & Exception Handling, Threads & Fibers Classes Modules & Objects: Simple Ruby Classes Object Instances							
III	Meta-programmi Strings, Variabl	es, Missing M	Types, Modules & Consta		6			
IV		ogram Structure:	names, declaratio variables, arrays, sli	n, variables, assignments,	6			
V	Data Types and	operations: , composite data	•	ontrol statements, methods,	6			
VI	Concurrency wi	Concurrency with Shared variables: ace condition, mutual exclusion, memory synchronization ,package						
			Text Books					
	avd Flanagan, Yuk Know", O'Reilly;		, "The Ruby Progra	amming Language: Everyth	ing You Need			
2 A		an, Brian W. K	Kernighan, "The G	Go Programming Language	e", Pearson			
			References					

1	Yukihiro Matsumoto, David Flanagan , "The Ruby Programming Language", Shroff,1st Edition, 2008.							
2	Caleb Doxsey, "An Introduction to Programming in Go", CreateSpace Independent Publishing Platform (3 September 2012)							
	Useful Links							
1	https://onlinecourses.swayam2.ac.in/aic20_sp37/preview							
2	https://www.javatpoint.com/ruby-tutorial							
3	https://www.ruby-lang.org/en/documentation/quickstart/							
4	https://gobyexample.com/							
5	https://www.javatpoint.com/go-tutorial							
6	https://www.coursera.org/specializations/google-golang							

CO-PO Mapping							
	Programme Outcomes (PO)						
	PO1	PO2	PO3	PO4	PO5	PO6	
CO1	3			1			
CO2		2				2	
CO3			2	2			

Assessment Plan based on Bloom's Taxonomy Level						
Bloom's Taxonomy Level	T1	T2	ESE	Total		
Remember	To be used minimum					
Understand	To be used	To be used	To be used	To be used		
	minimum	minimum	minimum	minimum		
Apply	10	10	20	40		
Analyze	5	5	15	25		
Evaluate	5	5	15	25		
Create			10	10		
Total	20	20	60	100		

				ege of Engineering,					
(Government Aided Autonomous Institute)									
AY 2021-22 Course Information									
Progra	Programme M.Tech. (CS and IT)								
Class,		ter	First Year M. Tech						
Cours			5IT513	, 50 1					
Cours			Professional Elect	ive – 1: Artificial Int	elligence				
Desire	d Requ	uisites:	Probability and Li						
		Scheme		Examination S					
Lectur		3 Hrs/week	<u>T1</u>	T2	ESE	Total			
Tutori		-	20	20	60	100			
Practi		-		<u> </u>					
Intera	ction	-		Cred	its: 3				
			Cou	rse Objectives					
1	To co	mnare variou		tificial Intelligence	<u> </u>				
2					s of Artificial Intell	igence			
3				Artificial Intelligen		1501100			
3	10 111			D) with Bloom's Tax					
At the	end of		students will be ab		onomy Level				
CO1			l concepts of Artif			Apply			
		•		onal structures of	Artificial	Analyse			
CO ₂		igence			1101110101				
CO3			stem in Artificial I	ntelligence		Create			
000	Buile	tun enpere sy		<u> </u>		Create			
Modu	le		Modu	le Contents		Hours			
	A	I and Problem	Solving by Search	1					
I				as state space search	, Uninformed search,	7			
			CSP problems						
II		nowledge Rep				7			
				ntation, First order lo	gıc-l				
III		nowledge Rea	O	t order logic-I, Baye	aion notresouls	6			
1111		ecision network		a order logic-i, Baye	sian network,	6			
		anning	<u> </u>						
IV			lanning. Plan space	planning, Planning	graph and	6			
		raphplan		r8,8	8				
		achine Learni	ing						
V	In	troduction to N	IL, Learning decision	on tress, Reinforcem	ent learning,	7			
			al network, Deep Lo	earning: A review.		/			
* **		xpert systems	· 1· /	CE .	A 1.4				
VI				ents of Expert syster	ns, Architecture of	6			
	E	S, Building an	Expert system						
			,	Text Books					
1	Rich	Elaine and Kel			e" McGraw Hills 3rd	edition 1991			
Janakiraman et al. "Foundations of Artificial Intelligence and Expert Systems" MacMillan India									
2 Ltd., 2007.									
References									
1		_	' Artificial Intellige	nce – A Modern App	roach", Prentice-Hall	, 2010 (3rd			
1	editio								
2			_	Shyamanta M Haza	rika , IIT Guwahati-"	Fundamentals			
	Of A	tificial Intellig	ence"						
Useful Links									
1		ıle I,II,III		oc19 me71/unit?unit	=781esson=8				

2	Module IV,V https://onlinecourses.nptel.ac.in/noc19_me71/unit?unit=16&lesson=17
3	Module VI Vlabs, iitb.ac.in

CO-PO Mapping								
Programme Outcomes (PO)								
	PO1	PO2	PO3	PO4	PO5	PO6		
CO1	1		3					
CO2		2				2		
CO3	2			1				

Assessment Plan based on Bloom's Taxonomy Level						
Bloom's Taxonomy Level	T1	T2	ESE	Total		
Remember	To be used minimum					
Understand	To be used minimum					
Apply	10	10	20	40		
Analyze	5	5	15	25		
Evaluate	5	5	15	25		
Create			10	10		
Total	20	20	60	100		

			Walchand College					
(Government Aided Autonomous Institute) AY 2021-22								
Course Information								
Progra	Programme M.Tech. (CS and IT)							
Class,		ter	First Year M. Tech					
Cours			5IT514	,				
Cours	e Nam	e	Professional Elect	ive - 2: Advanced I	Distributed Computing	<u> </u>		
Desire	d Requ	uisites:			, ,			
		g Scheme		Examination Sch	heme (Marks)			
Lectur		3 Hrs/week	T1	T2	ESE	Total		
Tutori		-	20	20	60	100		
Practi		-		~				
Intera	ction	-		Credit	ts: 3			
			C	01: 4:				
1	To :11	ustrate the veri		Objectives	ng			
2			us aspects of moder istributed architectu		115			
3			nd distributed comp					
3	1060		e Outcomes (CO) v		nomy Level			
At the	end of		students will be able		nomy Level			
CO1			various big data ana			Analyze		
CO2			roach to implement		nment	Analyze		
			y and performance			Evaluate		
CO ₃	syste		, 1	S				
Modu	le		Module	Contents		Hours		
			Distributed System			6		
I		Task Creation and Termination (Async, Finish), Tasks in Java's Fork/Join						
			putation Graphs, Work, Span, Multiprocessor Scheduling					
		Distributed System with Parallelism:						
II	I	Parallel Speedup, Amdahl's Law, Reciprocal Array Sum using Async- Finish,						
11	I	Reciprocal Array Sum using Recursive Action's in Java's Fork/Join						
		Framework						
		unctional Paral	llelism:					
III				utures in Java's Fork/Join Framework,		6		
			va Streams, Data Ra					
		•	ronization and Pip	O				
	1 1		ers with Java Phase	rs, Point-to-Point S	Synchronization with	7		
IV	I	Phasers,						
		One-Dimensional Iterative Averaging with Phasers, Pipeline Parallelism,						
		ata Flow Paralle						
		vistributed Map		on Framework Sa	ark Framowork TE			
V					ark Framework, TF- ge Rank Algorithm in	7		
		park	50 Kank Lampie, I	zemonsu auom. 1 ag	C Kank Aigoriann III	'		
		lient-Server Pr	ogramming:					
* 77			0	tion/Deserialization	n, Remote Method			
VI					Demonstration: File	6		
	Server using Sockets							
				t Books				
1					l Systems: Principles d	and		
			on, Pearson Educati		atuibated Cartana	lan acrete 1		
2	1	-	ean Dollimore, Ti Pearson Education	<u> </u>	stributed Systems: C	oncepts and		
			Rot	ferences				

1	A. S. Tanenbaum and M. V. Steen, "Distributed Systems: Principles and Paradigms", Second Edition, Prentice Hall, 2006
	Useful Links

Module I, II, III, IV

https://www.coursera.org/learn/parallel-programming-in-java?specialization=pcdp#syllabus

1 Module V, VI

https://www.coursera.org/learn/distributed-programming-in-

java?specialization=pcdp#syllabus

CO-PO Mapping							
Programme Outcomes (PO)							
	PO1	PO2	PO3	PO4	PO5	PO6	
CO1	3		1				
CO2		1			2		
CO3	1			2			

Assessment

Assessment Plan based on Bloom's Taxonomy Level						
Bloom's Taxonomy Level	T1	T2	ESE	Total		
Remember	To be used	To be used	To be used	To be used		
	minimum	minimum	minimum	minimum		
Understand	To be used	To be used	To be used	To be used		
	minimum	minimum	minimum	minimum		
Apply	10	10	20	40		
Analyze	5	5	15	25		
Evaluate	5	5	15	25		
Create			10	10		
Total	20	20	60	100		

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	(Government Aided Autonomous Institute) AY 2021-22								
Due ou	Programme M.Tech. (CS and IT)								
			M.Tech. (CS and						
Class,			First Year M. Tech	., Sem I					
Course				: 2. Ml A	liantian Davalannant				
			Web Technology	ive - 2: Modern App	lication Development				
Desire	u Kec	uisites:	web reciliology						
Та	achin	g Scheme		Examination So	chama (Marks)				
Lectur		3 Hrs/week	T1	T2	ESE	Total			
Tutori		J III S/ WEEK	20	20	60	100			
Practi		_	20	20	00	100			
Intera		_		Cred	its: 3				
mera	CUOII			Citu	113. 0				
			Cou	rse Objectives					
1	To d	emonstrate the s	static and dynamic w						
2			s for web using Scri	<u>, e e e e e e e e e e e e e e e e e e e</u>					
3			nsive web pages						
) with Bloom's Tax	onomy Level				
At the	end o	f the course, the	students will be abl	e to,	•				
CO1	Illus	trate the basic el	lements and propert	ies in different web a	applications.	Apply			
CO2	Deve	elop static and d	d dynamic web applications						
CO3	Desi	gn and develop	develop responsive web applications.						
Modu	le		Mod	ule Contents		Hours			
		HTML 5 and Bo	<u> </u>						
		Introduction, Getting Started, Grid System, Fixed Layout, Fluid Layout,							
_		Responsive Layout, Typography							
Ι		Bootstrap Basics Elements:							
	- 1	Jumbotron open link, Button, Button Groups, Grid, Table, Form, Alert, Wells,							
		Badge & Label, Panels, Pagination, Pager, Image, Glyphicon,, Carousel, Progress Bar, List Group, Dropdown, Collapse, Tabs.							
				, Collapse, Tabs.					
	- 1	ntroduction to		Madulas HTTDMa	41. IIDI M. 41.				
		Install Node.js Windows and Linux, Modules, HTTP Module, URL Module,							
II		First Example. Console, NPM: Package Manager, Node Globals, Node.js OS,							
11		Timer, Errors Node JS Basics:							
		Buffers, Streams, File System, Path, String Decoder, Query String, ZLIB,							
		Assertion, V8, Callbacks, Events, Punycode, TTY, Web Modules							
		Node JS and My		J , 1 1 1, 1, 20 1					
III				Create Table, Insert	Record, Update	6			
				d, Select Unique, Di					
	F	ReactJS:	<u> </u>		•				
	l I	ntroduction, Ten	nplating using JSX,	Components, State a	and Props, Lifecycle				
IV	О	f Components,	Rendering List and	Portals, Error Handli	ing, Routers, Redux	6			
1,	and Redux Saga, Immutable.js, Service Side Rendering, Unit Testing,								
	V	Vebpack							
	ID	ython Framew	ork ·						
				of Diango The Ra	sics of Dynamic, Wel	,			
V					Database: Models, The	、			
•				rm Processing, Fi		1 n			
			Sessions and Cookie	•	<u>-</u>				
		, ~				1			

VI	Ruby On Rails: Introduction, RVM(ruby version manager), Working in Linux(Ubuntu) Platform, Ruby Operators & Ruby Shell, Ruby Data types & Variables, Ruby methods and modules, OOP in Ruby, Basic loops and iterators. Rails: Rails Installation and Ruby gems, Databases, Statements, RAILS Model, Controller, and Views					
	Text Books					
1						
1	Benjamin Jakobus, "Mastering Bootstrap 4", Packt Publisher, 2nd Edition, 2018					
2	Jake Spurlock, "Bootstrap: Responsive Web Development", O'Reilly Publication, 1st Edition, 2013					
3	Ethan Brown, "Web Development using Node and Express", O'Really Publisher, 1st Edition, 2014.					
	References					
1	Daniel Rubio," Beginning Django Web Application Development and Deployment with Python ApressPublication,1st Edition,2017	,,,				
2	Michael Hartl," Ruby on Rails 3 Tutorial Learn Rails by Example", Pearson Education Publication, 1st Edition, 2010	on				
Useful Links						
1	https://www.tutorialsteacher.com/nodejs/nodejs-tutorials					
2	https://morioh.com/p/656c3d9c1bce					
3	https://www.tutorialrepublic.com/twitter-bootstrap-tutorial/					
4	https://morioh.com/p/11c3e757a913	\neg				
5	https://www.djangoproject.com/start/	\neg				

CO-PO Mapping								
Programme Outcomes (PO)								
	PO1	PO2	PO3	PO4	PO5	PO6		
CO1	2			2				
CO2			1		2			
CO3	1	3		1				

Assessment Plan based on Bloom's Taxonomy Level								
Bloom's Taxonomy Level	T1	T2	ESE	Total				
Remember	To be used	To be used	To be used	To be used				
	minimum	minimum	minimum	minimum				
Understand	To be used	To be used	To be used	To be used				
	minimum	minimum	minimum	minimum				
Apply	10	10	20	40				
Analyze	5	5	15	25				
Evaluate	5	5	15	25				
Create			10	10				
Total	20	20	60	100				

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			ided Autonomous Insti	itute)					
			AY 2021-22 rse Information						
Progran	Programme M.Tech. (CS and IT)								
	Semester	First Year M. Tech							
Course		5IT516	i., Sciii i						
Course			ive - 2: Image Process	sing and Pattern Reco	gnition				
	Requisites:		trix, Fourier Transform	<u> </u>	<u> Similon</u>				
		1							
Tea	ching Scheme		Examination Scl	heme (Marks)					
Lecture	2 3 Hrs/week	T1	T2	ESE	Total				
Tutoria	1 -	20	20	60	100				
Practica									
Interact	tion -		Credit	ts: 3					
a .	TD 1 1 1		rse Objectives						
	To apply mathematic								
	To compare image e		<u> </u>						
3	To elaborate image p		ons)) with Bloom's Taxo	momy Loyal					
At the a	and of the course, the			onomy Level					
			ssing for real-time ap	nlication	Apply				
	***			•	Apply				
	,	<u>1</u>			Analyze				
Module	e	Modu	le Contents		Hours				
I	Introduction and Pixel Relationship Need for Image Processing, Some Applications of Image Processing- Fundamental steps in DIP, Components of digital image processing, sampling, quantization, Pixel Relationships in images, Distance measurements, Data structure for image representation 1								
II	1 0 1		ons rations, Geometrical	operations , Image	7				
III	Image Transfor Need of transfor		operties, convolution	Theorem, DCT	6				
IV	•	,Spatial filtering tec	chniques, Frequency d	omain filtering	6				
V	Image Segmentation V Classification of Image segmentation, Edge detection, Thresholding techniques, Region growing techniques								
VI	Pattern Recognition Fundamentals Basic concepts of pattern recognition, fundamental problems in pattern								
1 :	S.Shridhar <i>"Dioital</i>		Text Books Oxford University Pr	ress.2 nd Edition 2016					
Millan sonka Vaclay Hiayac, Roger Royle, "Imaga Processing Analysis and Machine Vision" CI									
, ,	Engineering,3rd Edi								
1	S Jayraman S Feel		References nar, "Digital image pr	ocessina" 1st Edition	MGH 2017				
1			Digital Image Proces						
, ,	Education, 2008			, ora Danion, I					
		Į	Jseful Links						

1	Module I,II,III https://nptel.ac.in/courses/117/105/117105079/
2	Module IV,V https://nptel.ac.in/courses/106/105/106105223/
3	Module VI Vlabs.iitb.ac.in

CO-PO Mapping									
Programme Outcomes (PO)									
	PO1	PO2	PO3	PO4	PO5	PO6			
CO1	3			1					
CO2		2				2			
CO3	1		1						

Assessment Plan based on Bloom's Taxonomy Level								
Bloom's Taxonomy Level	T1	T2	ESE	Total				
Remember	To be used minimum							
Understand	To be used	To be used	To be used	To be used				
	minimum	minimum	minimum	minimum				
Apply	10	10	20	40				
Analyze	5	5	15	25				
Evaluate	5	5	15	25				
Create			10	10				
Total	20	20	60	100				

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			<u> </u>	ided Autonomous Ins. AY 2021-22	muie)	
				rse Information		
Progra	amma		M. Tech. (CS and			
Class,		ytom	First Year M. Tec	<u> </u>		
Cours			5IT521	II., Selli II		
Cours				hods and Application	2	
		uisites:	Data Willing Web	* *	8	
Desire	u Keq	uisites.	Database Eligiliee	aring		
Та	anchin	g Scheme		Examination So	homo (Marks)	
Lectur		3 Hrs/week	T1	T2	ESE	Total
Tutori		3 THS/ WEEK	20	20	60	100
Practi		<u>-</u>	20	20	00	100
Intera		<u>-</u>		Credi	ta. 2	
Intera	CHOII	-		Creui	<u>ıs: 3</u>	
			Con	ırse Objectives		
1	Too	vercise advance	d data mining tech			
2			algorithm for real-t			
3			solution for real w			
)) with Bloom's Tax	onomy Level	
At the			students will be ab			
CO1			es and algorithms			Apply
CO2		·		rithms for solving rea		Apply
CO3	Anal	yse various clu	stering and classific	cation techniques in	data mining	Analyse
Modu	Module Contents					
Introduction: Data Mining, Kinds of Data, Kinds of Patterns, Technologies, Major Issues in Data Mining. Getting to Know Your Data: Data Objects and Attribute Types, Basic Statistical Descriptions of Data, Data Visualization, Measuring Data Similarity and Dissimilarity						, 7
II		ata Pre-proces Data Cleaning, Pata Discretizati	Data Integration,	Data Reduction, Da	ta Transformation and	6
III	В	lining Frequer asic Concepts, lethods.		set Mining Method	ls, Pattern Evaluation	n 7
IV	Classification Racic Concepts Decision Tree Induction Rayes Classification Methods Rule-					
V	Cluster Analysis					
Outlier Detection Outliers and Outlier Analysis, Outlier Detection Methods, Statistical Approaches, Proximity-Based Approaches, Clustering-Based Approaches, Classification-Based Approaches, Mining Contextual and Collective Outliers, Outlier Detection in High-Dimensional Data						, 6
	ı			Text Books		
1					cepts and Techniques	The Morgan
	Kauf	mann Series in	Data Management	Systems ,3 rd Edition,	2011	dition 2002
2	unr	ын IVI. П, <i>D</i> а	ua mining. Introdi	iciory una Aavancea	topics", Pearson, 2 nd E	u111011, 2005
				References		

1	Chattamvelli Rajan, "Data Mining Methods: Concepts & Applications", Narosa Publishing House, 2 nd Edition, 2010				
2	Mitra Sushmita, Acharya Tinku, "Data Mining Multimedia, Soft Computing and Biometrics", WILEY Publication, 3rd Edition, 2003				
Useful Links					
1	https://onlinecourses.nptel.ac.in/noc20_cs12/preview				
2	https://www.javatpoint.com/data-mining				

CO-PO Mapping								
Programme Outcomes (PO)								
	PO1	PO2	PO3	PO4	PO5	PO6		
CO1	3				1			
CO2		2		2				
CO3	1		1					

Assessment Plan based on Bloom's Taxonomy Level							
Bloom's Taxonomy Level	T1	T2	ESE	Total			
Remember	To be used minimum						
Understand	To be used minimum						
Apply	10	10	20	40			
Analyze	5	5	15	25			
Evaluate	5	5	15	25			
Create			10	10			
Total	20	20	60	100			

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			`	Y 2021-22	insitiute)	
				se Information		
Progra	amme		M. Tech. (CS and			
Class,		ster	First Year M. Tech	<u> </u>		
Course			5IT522	, 501111		
Course			Scientific Comput	inσ		
		uisites:	Programming expe		Iava	
Desire	u req	uisites.	Trogramming empt	errence in e, e i i	, , , , , , , , , , , , , , , , , , , ,	
Te	eachin	g Scheme		Examination	Scheme (Marks)	
Lectur		3	T1	T2		Total
Tutori			20	20	60	100
Praction		_	20	20	00	100
Intera		_		Cr	redits: 3	
Intera	CHOII	-		CI	euits. 3	
			Com	rse Objectives		
1	To 11	se different prop	gramming paradigms	•	nuting	
2			programming langu			
3			rt writing using LA		ne prodein	
	10 4		rse Outcomes (CO		Saxonomy Level	
At the	end of		students will be able		unonomy never	
CO1			and logical program			Analyze
CO2			ming language for s		m	Apply
CO3			automate data form			Apply
		•		•		1 11 0
Modu	le		Mod	dule Contents		Hours
I	C	verview of the	Data Science and See Data Science pressification, Clustering	ocess, Scientific		7
II	N P	andas: Series,	ction, Numpy array,	naging missing	lexing, Numpy operations. data, groupby, merging &	6
III	P D G	ython for Data ata Visualizatio eographical Plo	Visualization: on through libraries lotting.	-	Seaborn, Plotly and Cufflinks,	6
IV	V		rs, Matrices, lists &		ogical vectored operators a using Factors in R	7
V	Image data type, Image representation, categorical data using Factors in R Data/Image Visualization using libraries — Using graphs to visualize data, Basic plotting in R, Manipulating the plotting window, Advanced plotting using lattice library in R. Image visualization in using 7					
VI	Image processing tools Data Reporting using LaTeX — VI LATEX Software installation, LATEX typesetting basics, LATEX math typesetting, Tables and matrices, Mathematics in Latex					
1		r Madhavan, "7 784390150		Text Books or Data Science"	, August 2015, Packt Publish	ing, ISBN:
				References		
1		-			ey-Cambridge Press, 5 th Editi	on, August
2	Gilbert Strang, "Introduction to linear algebra", Wellesley-Cambridge Press, 5 th Edition, August 2016 Douglas Montgomery, "Applied statistics and probability for engineers", 6 th Edition, Wiley					

Useful Links				
1	https://onlinecourses.nptel.ac.in/noc20_cs36/course			
2	https://spoken-tutorial.org/watch/Python+3.4.3/Plotting+Data/English/			

CO-PO Mapping						
		Programme Outcomes (PO)				
	PO1	PO2	PO3	PO4	PO5	PO6
CO1	2		2			1
CO2					2	
CO3		1	3			

Assessment Plan based on Bloom's Taxonomy Level						
Bloom's Taxonomy Level	T1	T2	ESE	Total		
Remember	To be used minimum					
Understand	To be used minimum					
Apply	10	10	20	40		
Analyze	5	5	15	25		
Evaluate	5	5	15	25		
Create			10	10		
Total	20	20	60	100		

	Walchand College of Engineering, Sangli				
	(Government Aided Autonomous Institute)				
	AY 2021-22				
	Course Information				
Programme	M. Tech. (CS and IT)				
Class, Semester	First Year M. Tech., Sem II				
Course Code	5IT571				
Course Name	Activity Based Lab for Data Mining Methods and Applications				
Desired Requisites:	Data Mining				

Teaching Scheme		Examination Scheme (Marks)				
Lecture	-	LA1	LA1 LA2 ESE Total			
Tutorial	-	30	30	40	100	
Practical	2 Hrs/Week					
Interaction	-	Credits: 1				

	Course Objectives				
1	To introduce student with concept of data mining				
2	To provide knowledge applications of Data Mining applications.				
3	To help students to address real-world challenges using Data mining algorithms.				
	Course Outcomes (CO) with Bloom's Taxonomy Level				
At the	At the end of the course, the students will be able to,				
CO1	Implement the software application using for data mining algorithm.	Apply			
CO2	Write & explain a detailed project report for submission and evaluation.	Evaluate			
CO3	Design and validate system for Data mining	Create			

List of Experiments / Lab Activities

List of Experiments:

Activities are to be carried out individually.

Each student will perform the activity based on course on following areas.

- 1. Design system for data analysis using data mining algorithms.
- 2. The system work on data set with different algorithm like classification, clustering, association, etc.
- 3. Industry Problem Statement(Sponsored Project)
- 4. Problem statements based on current or previously learned Technology.
- 5. At the end of the semester project group should achieve all the proposed objectives of the problem statement.
- 6. The work should be completed in all aspects of design, implementation and testing and follow software engineering practices.
- 7. Project report should be prepared and submitted in soft and hard form along with all the code and other dependency.

Student should perform the activities on the basis of the real-time applications in the subjects and submit the work with code, PPT, PDF, Text report document & reference material or on online GitHub.

Students should maintain activity log book containing weekly progress.

	5 6 51 6				
	Text Books				
1	Han Jiawei and Kamber Micheline "Data Mining - Concepts and Techniques" The Morgan				
	Kaufmann Series in Data Management Systems ,3rd Edition, , 2011				
2	Dunham M. H, "Data Mining: Introductory and Advanced topics", Pearson, 2ndEdition,				
	2003				
	References				
1	Chattamvelli Rajan, "Data Mining Methods: Concepts & Applications", Narosa Publishing				
1	House, 2 nd Edition, 2010				
2	Mitra Sushmita, Acharya Tinku, "Data Mining Multimedia, Soft Computing and Biometrics",				
	WILEY Publication, 3 rd Edition, 2003				

	Useful Links			
1	1 https://onlinecourses.nptel.ac.in/noc20_cs12/preview			
2	https://www.javatpoint.com/data-mining			

CO-PO Mapping						
		Programme Outcomes (PO)				
	PO1	PO2	PO3	PO4	PO5	PO6
CO1	1		2		1	
CO2		1				
CO3	3			3		

There are three components of lab assessment, LA1, LA2 and Lab ESE.

IMP: Lab ESE is a separate head of passing. LA1, LA2 together is treated as In-Semester Evaluation.

Assessment	Assessment Based on Conducted by Typical Schedule (for 26-week Sem)		Marks	
LA1	Lab activities,	Lab Course	During Week 1 to Week 6	30
attendance, journal Faculty Marks Su		Marks Submission at the end of Week 6	30	
LA2	Lab activities,	Lab Course	During Week 7 to Week 12	30
LAZ	attendance, journal	Faculty	Marks Submission at the end of Week 12	30
Lab ESE	Lab activities,	Lab Course	During Week 15 to Week 18	40
LauESE	attendance, journal	Faculty	Marks Submission at the end of Week 18	40

Assessment Plan based on Bloom's Taxonomy level					
Bloom's Taxonomy Level	LA1	LA2	Lab ESE	Total	
Remember	To be used minimum				
Understand	To be used minimum				
Apply	10	10	10	30	
Analyze	10	10	10	30	
Evaluate	5	5	10	20	
Create	5	5	10	20	
Total	30	30	40	100	

	Walchand College of Engineering, Sangli				
	(Government Aided Autonomous Institute)				
	AY 2021-22				
	Course Information				
Programme	M. Tech. (CS and IT)				
Class, Semester	First Year M. Tech., Sem II				
Course Code	5IT572				
Course Name	Activity Based Lab for Scientific Computing				
Desired Requisites:	Programming experience in C,C++,Java				

Teaching Scheme		Examination Scheme (Marks)					
Lecture	-	LA1	LA2	Lab ESE	Total		
Tutorial	-	30	30	40	100		
Practical	2 Hrs/Week						
Interaction	-	Credits: 1					

	Course Objectives					
1	To use different programming paradigms in scientific computing.					
2	To apply appropriate programming language for solving the problem					
3	To demonstrate report writing using LATEX tool.					
	Course Outcomes (CO) with Bloom's Taxonomy Level					
At the	end of the course, the students will be able to,					
CO1	Perform numerical computation using python libraries	Analyze				
CO2	Implement statistical computation using R libraries	Apply				
CO3	Compose the journal paper, reports using Open source tool (LATEX)	Create				

List of Experiments / Lab Activities

Activities:

2016

Activities are to be carried out individually.

Each student will perform the activity based on course on following areas.

- 1. Exercise programs on Lists.
- 2. Exercise programs on Tuples.
- 3. Exercise programs on sets and dictionaries
- 4. Exercise programs on files.
 - a) Write Python script to display file contents.
 - b) Write Python script to copy file contents from one file to another.
- 5. Data visualization plots in R
- 6. Exercise programs on Vectors, Matrices, lists in R
- 7. Exercise programs on Data frames and factors in R
- 8. Exercise program on image libraries using R
- 9. Create a journal paper using open source tool LATEX
- 10. Create a seminar/project report using open source tool LATEX

Student should perform the activities on the basis of the real-time applications in the subjects and submit the work with code, PPT, PDF, Text report document & reference material or on online GitHub. Students should maintain activity log book containing weekly progress.

	Text Books				
1	Douglas Montgomery, "Applied statistics and probability for engineers", 6th Edition, Wiley				
1	Publications, January 2016				
2	Samir Madhavan, "Mastering Python for Data Science", August 2015, Packt Publishing, ISBN:				
	9781784390150				
References					
1	Gilbert Strang, "Introduction to linear algebra", Wellesley-Cambridge Press, 5th Edition, August				

	Useful Links					
1	https://docs.python.org					
2	https://www.docs.rstudio.com					
3	https://www.overleaf.com					

CO-PO Mapping							
		Programme Outcomes (PO)					
	PO1	PO2	PO3	PO4	PO5	PO6	
CO1	3			3			
CO2							
CO3	1	2	2			1	

There are three components of lab assessment, LA1, LA2 and Lab ESE.

IMP: Lab ESE is a separate head of passing. LA1, LA2 together is treated as In-Semester Evaluation.

Assessment	Based on	Conducted by	Typical Schedule (for 26-week Sem)	Marks
LA1	Lab activities,	Lab Course	During Week 1 to Week 6	30
LAI	attendance, journal	Faculty	Marks Submission at the end of Week 6	30
LA2	Lab activities, Lab Course During Week 7 to Week 12		During Week 7 to Week 12	30
	attendance, journal	Faculty	Marks Submission at the end of Week 12	30
Lab ESE	Lab activities,	Lab Course	During Week 15 to Week 18	40
	attendance, journal	Faculty	Marks Submission at the end of Week 18	40

Assessment Plan based on Bloom's Taxonomy level							
Bloom's Taxonomy Level	LA1	LA2	Lab ESE	Total			
Remember	To be used minimum						
Understand	To be used minimum						
Apply	10	10	10	30			
Analyze	10	10	10	30			
Evaluate	5	5	10	20			
Create	5	5	10	20			
Total	30	30	40	100			

Walchand College of Engineering, Sangli (Government Aided Autonomous Institute) AY 2021-22 Course Information Programme M. Tech. (CS and IT) Class, Semester First Year M. Tech., Sem II Course Code 5IT573 Course Name Industrial Project Desired Requisites:

Teachin	g Scheme	Examination Scheme (Marks)				
Lecture	-	LA1	LA2	Lab ESE	Total	
Tutorial	-	30	30	40	100	
Practical	-					
Interaction	2 Hrs/Week	Credits: 2				

	Course Objectives					
1	To elaborate the recent methods and trends in industrial projects					
2	To exercise the industry practices and standards					
3	To explore the industry ready applications and projects					
	Course Outcomes (CO) with Bloom's Taxonomy Level					
At the	At the end of the course, the students will be able to,					
CO1	Analyze various aspects, methodologies and practices in industry	Analyze				
CO2	Demonstrate the trends in industry	Apply				
CO3	Design industry ready applications	Create				

List of Experiments / Lab Activities

Activities:

Activities are to be carried out individually.

Each student will perform the activity based on course on following areas.

Topic Selection:

Significance and Scope of comprehensive topic with exploration at each level, technical competency with Research oriented topic, literature survey of reliable and valid sources. Responsibly summarized literature

Relevance to Dissertation:

At least three topics in relevance to thirst area of dissertation need to be overlooked.

Scope of Topic:

Relevance, significance and expected outcome discussion in stated problem statements for area of dissertations.

Report writing:

Proper citation of sources, organized section of chapters, standard and valid references, nearly absolute contents.

This course will include carrying out a project considering the social needs, innovative designing, and implementation as well as exploring its commercialization / patenting of the project.

Visit nearby industry for selection of problem statement

Text Books	
1	
References	
1	
Useful Links	
1	

	Programme Outcomes (PO)					
	PO1	PO2	PO3	PO4	PO5	PO6
CO1	1		1			
CO2	2			2	1	
CO3		3		1		

There are three components of lab assessment, LA1, LA2 and Lab ESE.

IMP: Lab ESE is a separate head of passing. LA1, LA2 together is treated as In-Semester Evaluation.

Assessment	Based on	Conducted by	Typical Schedule (for 26-week Sem)	Marks
T A 1	Lab activities,	Lab Course	During Week 1 to Week 6	20
LA1	attendance, journal	Faculty	Marks Submission at the end of Week 6	30
LA2	Lab activities, Lab Course		During Week 7 to Week 12	30
LAZ	attendance, journal	Faculty	Marks Submission at the end of Week 12	30
Lab ECE	Lab activities,	Lab Course	During Week 15 to Week 18	40
Lab ESE	attendance, journal	Faculty	Marks Submission at the end of Week 18	40

Assessment Plan based on Bloom's Taxonomy level				
Bloom's Taxonomy Level	LA1	LA2	Lab ESE	Total
Remember	To be used minimum			
Understand	To be used minimum			
Apply	10	10	10	30
Analyze	10	10	10	30
Evaluate	5	5	10	20
Create	5	5	10	20
Total	30	30	40	100

			Walchand Co	llege of Engineerin	g, Sangli					
	(Government Aided Autonomous Institute)									
	AY 2021-22									
_			1	urse Information						
Progra			M. Tech. (CS and	· · · · · · · · · · · · · · · · · · ·						
Class,			First Year M. Tech	ı., Sem II						
Course			5IT523	ivo 2. Distributed	On anoting Creatons					
				ive – 3: Distributed (s, Distributed Netwo						
Desire	u Keq	uisites:	Operating Systems	s, Distributed Netwo	1K					
То	achin	g Schome		Evamination	Scheme (Marks)					
Teaching Scheme Lecture 2 Hrs/week			T1	T2	ESE	Total				
Tutori		-	20	20	60	100				
Practio		-								
Intera		-		Cro	edits: 2					
			ı							
			Co	ourse Objectives						
1	To el	aborate fundar	nental characteristic	s of distributed oper	ating systems.					
2			ited operating syster							
3	To in	terpret the co	mmunication, proce	ss, naming, synchro	nization in distribute	ed operating systems				
		C	ourse Outcomes (C	CO) with Bloom's T	axonomy Level					
At the	end of	the course, the	e students will be ab	le to,						
CO1	Anal	yze the charac	teristics of distribut	ed operating system	S	Analyze				
CO2	Use	distributed oper	rating systems for di	1	Apply					
CO3	Com	pare various di	stributed operating s	systems		Analyze				
Modu				Contents		Hours				
I			distributed System			4				
				Software concepts,		·				
П	Communication & Synchronization in distributed systems: Computer Network and Layered protocols, Message passing and related issues, synchronization, Client Server model & its implementation, remote procedure call and implementation issues, Case Studies: SUN RPC, DEC RPC Clock synchronization and related algorithms, mutual exclusion, Deadlock in distributed systems									
III	P T sy	rocesses and phreads, system to Load busteributed systems:	processors & Distri n model, processor palancing and sharin	buted File Systems allocation, schedul g approach, fault to on and related issues ributed file system,	ing in distributed lerance, Real time	4				
IV	Ir ir	ntroduction, g nplementation	issues of DSM, gra	e of DSM syste anularity, structure on nent strategy, thrashi	of shared memory	4				
V	space, consistency models, replacement strategy, thrashing Naming & Distributed Web-based Systems: Overview, Features, Basic concepts, System oriented names, Object locating mechanisms, Issues in designing human oriented names, Name caches, Naming and security, DNS Architecture, Processes, Communication, Naming, Synchronization, Consistency and Replication: Web Proxy Caching, Replication for Web Hosting Systems, Replication of Web Applications									
VI	Security & Case Study: Google ES/RigTable Introduction of Security in Distributed OS									
1		eep K. Sinha ".	Distributed Operation	Text Books ng Systems Concepts	s and Design", Easte	ern Economy Edition,				
	1111, 1990.									

2	George Coulouris, Jean Dollimore, Tim Kindberg "Distributed Systems: Concepts and Design", Fifth Edition, Pearson, 2012.						
	References						
1	Sunita Mahajan & Seema Shah, "Distributed Computing", Second Edition, OXFORD, 2013						
	Useful Links						
1	https://nptel.ac.in/courses/106/106/106106107/						
2	https://nptel.ac.in/courses/106/106/106106168/						

CO-PO Mapping							
	Programme Outcomes (PO)						
	PO1	PO2	PO3	PO4	PO5	PO6	
CO1	1		3			2	
CO2		1		2			
CO3	2		1				

Assessment Plan based on Bloom's Taxonomy Level									
Bloom's Taxonomy Level	T1	T2	ESE	Total					
Remember	To be used	To be used	To be used	To be used					
	minimum	minimum	minimum	minimum					
Understand	To be used	To be used	To be used	To be used					
	minimum	minimum	minimum	minimum					
Apply	10	10	20	40					
Analyze	5	5	15	25					
Evaluate	5	5	15	25					
Create			10	10					
Total	20	20	60	100					

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				Y 2021-22				
D				se Information				
	amme		M. Tech. (CS and	<u> </u>				
	Semes		First Year M. Tecl	n., Sem II				
	se Code		5IT524		•			
	se Nam			ive – 3: System Prog				
Desire	ed Req	uisites:	Data Structures an	d Operating Systems	S			
Т	aaahin	g Scheme		Examination So	phomo (Morks)			
Lectu		2 Hrs/week	T1	T2	ESE	т	otal	
Tutor		Z HIS/WEEK	20	20	60		100	
		-	20	20	00		100	
Practi		-		C 1	· · · · · · · · · · · · · · · · · · ·			
Intera	iction	-		Credi	its: 2			
			Com	rse Objectives				
1	To al	aborate the con	cepts in systems pro					
2			<u> </u>	ssemblers, linkers an	d loaders			
3				nguages executions u		ms		
) with Bloom's Tax				
At the			students will be able					
CO1		Analyze the working of system programs Analyze the working of system programs						
CO2			of parsers of compilers					
CO3 Compare the static and dynamic linking						Analyze		
Modu	ıle		Mod	dule Contents			Hours	
1120020		verview of Sys					110011	
I	Ir S S P A	ntroduction, So tructure, Interfactorice Program, rogramming I activities, Program	oftware, Software cases, Address Space, Levels of System Languages and Languages	Hierarchy, Systems, Computer Languag Software, Overview anguage Processor ndamental of Lang	ges, Tools, Life Cyo v of Language Pro s, Language Pro	cle of a cessors cessing	4	
II	A E A A	Activities, Program Execution, Fundamental of Language Processing, Symbol Tables Assemblers: Elements of Assembly Language Programming, Design of the Assembler, Assembler Design Criteria, Types of Assemblers, Two-Pass Assemblers, One-Pass Assemblers, Single pass Assembler for Intel x86, Algorithm of Single Pass Assembler, Multi-Pass Assemblers, Advanced Assembly Process, Variants of Assemblers Design of two pass assemblers.						
III	Ir A A D T	Assemblers Design of two pass assembler, Macro and Macro Processors: Introduction, Macro Definition and Call, Macro Expansion, Nested Macro Calls, Advanced Macro Facilities, Design Of a Macro Pre-processor, Design of a Macro Assembler, Functions of a Macro Processor, Basic Tasks of a Macro Processor, Design Issues of Macro Processors, Features, Macro Processor Design Options, Two-Pass Macro Processors, One-Pass Macro Processors						
IV	Ir P L C L	rograms, Linkir inking, Loader ompile-and-Go	location of Linking ng in MSDOS, Link is, Different Loadir Loaders, General L cal Relocating Loa	Concept, Design of ing of Overlay Struc- ng Schemes, Sequen- coader Schemes, Absolders, Linking Load	ctured Programs, D ntial and Direct L solute Loaders, Rel	ynamic oaders, ocating	5	

Loaders, Linkers v/s Loaders

V	Scanning and Parsing: Programming Language Grammars, Classification of Grammar, Ambiguity in Grammatic Specification, Scanning, Parsing, Top Down Parsing, Bottom up Parsing, Language Processor Development Tools, LEX, YACC, Compilers: Causes of Large Semantic Gap, Binding and Binding Times, Data Structure used in Compiling, Scope Rules, Memory Allocation, Compilation of Expression, Compilation of Control Structure, Code Optimization	4				
VI	VI Interpreters & Debuggers: Benefits of Interpretation, Overview of Interpretation, The Java Language Environment, Java Virtual Machine, Types of Errors, Debugging Procedures, Classification of Debuggers, Dynamic/Interactive Debugger					
	Text Books					
	D M Dhamdhere, "System Programming", McGraw Hill Publication, second revise	d edition				
1	1 2009					
2	2 Srimanta Pal, "System Programming", Oxford University Press, 2011					
3	D.V. Mayerya & A. Godhala "System Programming and Compiler Construction" Droomtoch					
	References					
1	Leland L. Beck, "System Software – An Introduction to Systems Programming" Education Asia,3 rd edition, 2000	, Pearson				
2	Santanu Chattopadhyay, 'System Software", Prentice-Hall India, 2007					
3	D. K. Maurya and Anand A. Godhala "System Programming and Compiler Construction (Includes					
	Useful Links					
1	www.cs.jhu.edu/~scott/pl/lectures/parsing.html					
2	www.en.wikipedia.org/wiki/System_programming					
3	https://nptel.ac.in/courses/106/106/106106197/					

CO-PO Mapping							
	Programme Outcomes (PO)						
	PO1	PO2	PO3	PO4	PO5	PO6	
CO1	2		3				
CO2	1	2		1			
CO3			1			1	

Assessment Plan based on Bloom's Taxonomy Level								
Bloom's Taxonomy Level	T1	T2	ESE	Total				
Remember	To be used	To be used	To be used	To be used				
	minimum	minimum	minimum	minimum				
Understand	To be used	To be used	To be used	To be used				
	minimum	minimum	minimum	minimum				
Apply	10	10	20	40				
Analyze	5	5	15	25				
Evaluate	5	5	15	25				
Create			10	10				
Total	20	20	60	100				

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			AY 2021-22					
			rse Information					
Progra		M. Tech. (CS and						
	Semester	First Year M. Tec	h., Sem II					
	Code	5IT525	· 2 M 4	C M 1' T '				
	e Name		ive - 3: Mathematics	for Machine Learning				
Desire	d Requisites:	Mathematics						
То	aching Scheme		Examination Sc	homo (Marks)				
Lectur	Ÿ	T1	T2	ESE	Total			
Cutori	-	20	20	60	100			
Practio			20	00	100			
nterac			Credi	ts: 2				
		Cou	rse Objectives					
1	To use linear algebra	and calculus for n	nachine learning.					
2	To elaborate matrix	theory for machine	learning.					
3		<u> </u>	ty for real application	S				
'	Cou	rse Outcomes (CO)) with Bloom's Tax	onomy Level				
At the	end of the course, the							
CO1	Apply the concepts of	of linear algebra an	d calculus for machin	e learning algorithms	Apply			
	Q 11.00							
CO2	Compare different al	<u> </u>	<u> </u>		Analyse			
CO3	Evaluate the optimiz	Evaluate the optimization & probabilistic algorithms						
					Hours			
Modu								
I	Linear Algebra Vector spaces ar		and dimensions, lin	ear transformation, four	4			
_	Vector spaces and subspaces, basis and dimensions, linear transformation, four fundamental subspaces.							
	Matrix Theory:							
	Norms and spaces, eigenvalues and eigenvectors, Special Matrices and their							
II	_	properties, least squared and minimum normed solutions. SVD: Properties and						
	applications, low rank approximations, Gram Schmidt process, polar							
	decomposition							
		luction Algorithm	s:					
Dimensions Reduction Algorithms: III Principal component analysis, linear discriminant analysis, minimal polynon					4			
III	and Jordan canonical form.							
III	Calculus:	iicur romii						
	Calculus:		derivatives, gradient.	directional derivatives.	4			
III	Calculus: Basic concepts of	of calculus: partial		directional derivatives,	4			
	Calculus: Basic concepts of	of calculus: partial	derivatives, gradient, vex functions and its J		4			
	Calculus: Basic concepts of Jacobian, hessian Optimization:	of calculus: partial	vex functions and its J	properties.				
	Calculus: Basic concepts of Jacobian, hessian Optimization: Unconstrained an	of calculus: partial a, convex sets, convex and Constrained opti	vex functions and its j imization, Numerical	oroperties. optimization techniques				
IV	Calculus: Basic concepts of Jacobian, hessian Optimization: Unconstrained ar for constrained	of calculus: partial a, convex sets, convex and Constrained option	vex functions and its jumination, Numerical displaying optimization: New	properties.	5			
IV	Calculus: Basic concepts of Jacobian, hessian Optimization: Unconstrained ar for constrained descent method	of calculus: partial a, convex sets, convex and Constrained opti and unconstrained b, Penalty function	vex functions and its jumination, Numerical displaying optimization: New	optimization techniques ton's method, Steepest ction to SVM, Error	5			
IV	Calculus: Basic concepts of Jacobian, hessian Optimization: Unconstrained ar for constrained descent method	of calculus: partial a, convex sets, convex and Constrained opti and unconstrained b, Penalty function	vex functions and its jumization, Numerical optimization: New on method. Introdu	optimization techniques ton's method, Steepest ction to SVM, Error	5			
IV V	Calculus: Basic concepts of Jacobian, hessian Optimization: Unconstrained ar for constrained descent method minimizing LPP, Probability: Basic concepts	of calculus: partial a, convex sets, convex sets, convex and Constrained option and unconstrained, Penalty function concepts of duality of probability:	imization, Numerical optimization: New on method. Introduy, hard and soft marg	optimization techniques ton's method, Steepest ction to SVM, Error in classifiers.	5			
IV	Calculus: Basic concepts of Jacobian, hessiar Optimization: Unconstrained an for constrained descent method minimizing LPP, Probability: Basic concepts independence, the	of calculus: partial a, convex sets, convex	imization, Numerical optimization: New on method. Introdu y, hard and soft marg conditional probability, expectat	optimization techniques ton's method, Steepest ction to SVM, Error in classifiers. lity, Bayes' theorem, ion and variance, few	5			
IV V	Calculus: Basic concepts of Jacobian, hessiar Optimization: Unconstrained an for constrained descent method minimizing LPP, Probability: Basic concepts independence, the	of calculus: partial a, convex sets, convex	imization, Numerical optimization: New on method. Introduy, hard and soft marg	optimization techniques ton's method, Steepest ction to SVM, Error in classifiers. lity, Bayes' theorem, ion and variance, few	5			
IV V	Calculus: Basic concepts of Jacobian, hessiar Optimization: Unconstrained an for constrained descent method minimizing LPP, Probability: Basic concepts independence, the	of calculus: partial a, convex sets, convex sets, convex and Constrained option and unconstrained, Penalty function concepts of duality of probability: theorem of total pinuous distribution	imization, Numerical optimization: New on method. Introdu y, hard and soft marg conditional probably probability, expectat s, joint distributions a	optimization techniques ton's method, Steepest ction to SVM, Error in classifiers. lity, Bayes' theorem, ion and variance, few	5			
IV V	Calculus: Basic concepts of Jacobian, hessian Optimization: Unconstrained ar for constrained descent method minimizing LPP, Probability: Basic concepts independence, tildiscrete and contents	of calculus: partial a, convex sets, convex sets, convex and Constrained option and unconstrained, Penalty function concepts of duality of probability: heorem of total jinuous distribution	imization, Numerical optimization: New on method. Introdu y, hard and soft marg conditional probability, expectat s, joint distributions a	optimization techniques ton's method, Steepest ction to SVM, Error in classifiers. lity, Bayes' theorem, ion and variance, few and co-variance.	5 4			
IV V	Calculus: Basic concepts of Jacobian, hessian Optimization: Unconstrained ar for constrained descent method minimizing LPP, Probability: Basic concepts independence, tildiscrete and contents	of calculus: partial a, convex sets, convex sets, convex and Constrained option and unconstrained, Penalty function concepts of duality of probability: heorem of total jinuous distribution	imization, Numerical optimization: New on method. Introdu y, hard and soft marg conditional probability, expectat s, joint distributions a	optimization techniques ton's method, Steepest ction to SVM, Error in classifiers. lity, Bayes' theorem, ion and variance, few	5 4			

	References					
1	All Modules taken from below link course.					
1	https://onlinecourses.nptel.ac.in/noc21_ma38/					
	Useful Links					
1	https://nptel.ac.in/courses/111/107/111107137/					

CO-PO Mapping							
		Programme Outcomes (PO)					
	PO1	PO2	PO3	PO4	PO5	PO6	
CO1	2			2			
CO2		3	3		2		
CO3	1			1			

Assessment Plan based on Bloom's Taxonomy Level								
Bloom's Taxonomy Level	T1	T2	ESE	Total				
Remember	To be used minimum							
Understand	To be used minimum							
Apply	10	10	20	40				
Analyze	5	5	15	25				
Evaluate	5	5	15	25				
Create			10	10				
Total	20	20	60	100				

				of Engineering, San					
	(Government Aided Autonomous Institute)								
	AY 2021-22 Course Information								
Program	Programme M. Tech. (CS and IT)								
	Class, Semester First Year M. Tech., Sem II								
	Course Code 5IT526								
Course				e - 4: Big Data Comp	uitino				
Desired			Data Mining	c 1. Dig Duta Comp	umg				
Desired	. rioqu		2						
Te	eaching	g Scheme		Examination Sche	eme (Marks)				
Lecture		2 Hrs/week	T1	T2	ESE	Total			
Tutoria	al	-	20	20	60	100			
Practic	al	-							
Interac	tion	-		Credits	: 2				
				Objectives					
1			mental concepts of b						
2			a using various techn						
3	To re		using visualization to						
				ith Bloom's Taxono	my Level				
			udents will be able to						
CO1				lata analytics techniqu		Apply			
CO2				stributed environmen		Analyze			
CO3	Evalu	ate the performa	nce of algorithms on	advanced distributed	system	Evaluate			
Modul				e Contents		Hours			
I	B In	troduction to Big	nportance, Four V's og Data Analytics, Big	of Big Data, Drivers f Data Analytics appli	•	4			
II	H A	nalytics, Cloud a	World, Data discovered Big Data, Predict	• •	hnology for Big Data Business Intelligence	5			
III	Pi D Id cu ar	rocessing Big Da etecting Patterns lentifying previous stomer market v propriate distance	ata: in Complex Data wiusly unknown groupivith the K-Means algo measures, Constru	th Clustering and Linings within a data set, orithm, Defining similating tree—like cluster and tweets to aid under	k Analysis, Segmenting the clarity with s with hierarchical	4			
IV		adoop Mapredu		ramework, Spark Fra	mework	5			
V	D T	istributed Map	Reduce:	•	ge Rank Algorithm in	4			
VI	Analytic Tools:								
			TET.	4 D. I.					
1	D'	1		t Books	o alse Deale Hall and a 18t ID 11	diam 2012			
2	Mine	lli Michael, Cha	mbers Michehe, "Bi	g Data, Big Analytic	ackt Publishing, 1 st Edi s: Emerging Business ely CIO Series, 1st Editi	Intelligence			
			Ref	erences					
1	References 1 Franks Bill, "Taming the Big Data Tidal Wave: Finding Opportunities in Huge Data Streams with Advanced Analytics", Wiley and SAS Business Series,1st Edition, 2012								
			Usef	ul Links					

CO-PO Mapping								
		Programme Outcomes (PO)						
	PO1	PO2	PO3	PO4	PO5	PO6		
CO1	1		1					
CO2		1		2				
CO3	3		2		1			

Assessment Plan based on Bloom's Taxonomy Level								
Bloom's Taxonomy Level	T1	T2	ESE	Total				
Remember	To be used	To be used	To be used	To be used				
	minimum	minimum	minimum	minimum				
Understand	To be used	To be used	To be used	To be used				
	minimum	minimum	minimum	minimum				
Apply	10	10	20	40				
Analyze	5	5	15	25				
Evaluate	5	5	15	25				
Create			10	10				
Total	20	20	60	100				

				ege of Engineering		
			· · · · · · · · · · · · · · · · · · ·	Y 2021-22	nstitute)	
				se Information		
Progra	mme		M. Tech. (CS and	IT)		
	Semester		First Year M. Tec	h., Sem II		
Course			5IT527			
Course					ormance Computing	
Desired	d Requisit	tes:	Operating System			
	aching Sc			Examination	Scheme (Marks)	
Lectur		Irs/week	T1	T2		Total
Tutoria	-	-	20	20	60	100
Practic Interac		-		C	edits: 2	
mterac	:11011			Cr	euits: 2	
			Cou	rse Objectives		
1					rformance computing	
2			ormance of parallel		quential program	
3	To compa	are multi-co	ore and many-core a	rchitectures		
		Con	rse Outcomes (CO) with Bloom's T	axonomy Level	
At the	end of the		students will be abl	<i>'</i>	axonomy Ecver	
CO1			uting algorithm for		m.	Apply
CO2			implemented algori			Analyze
CO3	Design th	ne appropria	te parallel algorithr	n for the given pro	blem.	Create
Modul	e		Mo	dule Contents		Hours
		Parallel A				
I	Comp		Parallel, Monte Car		on the JVM, Running mate Pi, First-Class Tasks	5
II	Parall	el Sorting,	•		oing, Parallel Fold (Reduce)	4
III	Data- Parall	Parallelisn el Collectio	ons		n-Parallel Operations, Scala	5
T	Data		for Parallel Comp	0		
IV		ure, Amort			onstruction, Conc-tree Data n, Conc-Tree Combiners	4
V	Issues	0	twork, Bubble sort			4
VI	_	SSSP, APS				4
			7	Text Books		
1				rypis, Vipin Kum	ar, "Introduction to parallel	computing,
				References		
1					Computer Science, W. H. F.	reeman and
			T1	seful Links		
		, II, III, IV			0EAID	ranSiteID
1		ww.courser	a.org/learn/parprog	1 ?ranivi1D=403286	&ranEAID=*GqSdLGGurk&	Tansicis

utm_campaign=*GqSdLGGurk#syllabus

CO-PO Mapping							
	Programme Outcomes (PO)						
	PO1	PO2	PO3	PO4	PO5	PO6	
CO1		1		1			
CO2	2		2		1		
CO3		2		3			

Assessment

Assessment Plan based on Bloom's Taxonomy Level							
Bloom's Taxonomy Level	T1	Т2	ESE	Total			
Remember	To be used	To be used	To be used	To be used			
	minimum	minimum	minimum	minimum			
Understand	To be used	To be used	To be used	To be used			
	minimum	minimum	minimum	minimum			
Apply	10	10	20	40			
Analyze	5	5	15	25			
Evaluate	5	5	15	25			
Create			10	10			
Total	20	20	60	100			

		A	Y 2021-22						
		Cour	se Information						
Programn	ne	M. Tech. (CS and	IT)						
Class, Semester First Year M. Tech., Sem II									
Course Co	ode	5IT528	·						
Course Na	ame	Professional Elect	ive - 4: Deep Learnir	ng					
Desired R	equisites:		•						
	-	1							
Teach	ing Scheme		Examination Sc	cheme (Marks)					
Lecture	2 Hrs/week	T1	T2	ESE	7	Γotal			
Tutorial	-	20	20	60		100			
Practical	-								
Interactio	n -		Credi	ts: 2					
	'								
		Cou	rse Objectives						
1 To	elaborate the mo	dels of Deep Learnin	ng						
			arning with performa	ance parameters					
3 To	interpret the prob	olem to solve using l	Deep Learning						
A 1		,) with Bloom's Tax	onomy Level					
	<u> </u>	students will be abl	<u> </u>	··		A 1			
	<u> </u>		ng for suitable applic			Apply			
				npare the optimization techniques pertaining to Deep Learning Analyze					
CO3 Build and compare various Deep Learning model for solving real world application Create									
	•	various Deep Learn	ing model for solving	g real world applicati	on	Create			
Module		•		g real world applicati	on				
Module	Fundamentals	Mod	ule Contents	g real world applicati	on	Hours			
Module		Mod f Neural Networks	lule Contents						
	McCulloch Pitts	Mod of Neural Networks Neuron, Threshold	lule Contents : ing Logic, Perceptro	ons, Perceptron Lear	rning	Hours			
Module I	McCulloch Pitts Algorithm, Mul	Mod f Neural Networks Neuron, Threshold tilayer Perceptrons	ing Logic, Perceptro (MLPs), Represent	ons, Perceptron Lear	rning LPs,				
	McCulloch Pitts Algorithm, Mul Sigmoid Neuro	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D	ing Logic, Perceptro (MLPs), Representescent, Feedforwa	ons, Perceptron Lear tation Power of M	rning LPs, orks,	Hours			
	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm.	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor	ing Logic, Perceptro (MLPs), Representescent, Feedforward Neural Netv	ons, Perceptron Lear tation Power of M rd Neural Netwo	rning LPs, orks,	Hours			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor	ing Logic, Perceptro (MLPs), Representescent, Feedforward Neural Networks	ons, Perceptron Lear tation Power of M rd Neural Netwo works. Backpropaga	ning LPs, orks, ation	Hours			
	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descent	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent ont (GD), Momentu	ing Logic, Perceptro (MLPs), Representescent, Feedforward Neural Networks Neural Ne	ons, Perceptron Lear tation Power of M rd Neural Netwo works. Backpropaga	ning LPs, orks, ation	Hours			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descention	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent ont (GD), Momentu AdaGrad, RMSProp,	ing Logic, Perceptro (MLPs), Representescent, Feedforward Neural Networks	ons, Perceptron Lear tation Power of M rd Neural Netwo works. Backpropaga	ning LPs, orks, ation	Hours 5			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descention Stochastic GD, A Regularization:	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent nt (GD), Momentu AdaGrad, RMSProp,	ing Logic, Perceptro (MLPs), Represent escent, Feedforwarward Neural Networks.	ons, Perceptron Lear tation Power of M rd Neural Netwo works. Backpropaga sterov Accelerated on in Adam.	rning LPs, orks, ation	Hours 5			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descention Stochastic GD, A Regularization: Regularization:	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent ont (GD), Momentu AdaGrad, RMSProp, Bias Variance Trade	ing Logic, Perceptro (MLPs), Representescent, Feedforward Neural Networks and Based GD, New Adam, Bais correction off, L2 regularization	ons, Perceptron Lear tation Power of M rd Neural Netwoworks. Backpropagasterov Accelerated on in Adam.	ming LPs, orks, ation GD,	Hours 5			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descent Stochastic GD, A Regularization: Regularization: set augmentation	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent ont (GD), Momentu AdaGrad, RMSProp, Bias Variance Trade of Parameter sharing	ing Logic, Perceptro (MLPs), Representescent, Feedforwar ward Neural Network Based GD, New Adam, Bais correction off, L2 regularization and tying, Injecting	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam.	rning LPs, orks, ation GD,	Hours 5			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descen Stochastic GD, A Regularization: Regularization: I set augmentation methods, Droport	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent ont (GD), Momenta AdaGrad, RMSProp, Bias Variance Trade on, Parameter sharing out. Greedy Layer w	ing Logic, Perceptro (MLPs), Representescent, Feedforwar ward Neural Network Adam, Bais correction off, L2 regularization and tying, Injecting vise Pre-training, Be	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, Enoise at input, Ensertter activation functions.	rning LPs, orks, ation GD,	Hours 5			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descention Stochastic GD, A Regularization: Regularization: Set augmentation methods, Dropost Better weight ini	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent on (GD), Momentu AdaGrad, RMSProp, Bias Variance Trade on, Parameter sharing out. Greedy Layer w tialization methods,	ing Logic, Perceptro (MLPs), Represent escent, Feedforwar ward Neural Network Adam, Bais correction off, L2 regularization and tying, Injecting vise Pre-training, Be Batch Normalization	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, Enoise at input, Ensertter activation function.	rning LPs, orks, ation GD,	Hours 5			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descent Stochastic GD, A Regularization: Regularization: set augmentation methods, Dropot Better weight ini Deep Learning for	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent on (GD), Momenta AdaGrad, RMSProp, Bias Variance Trade on, Parameter sharing out. Greedy Layer w tialization methods, for word encoding-	ing Logic, Perceptro (MLPs), Representescent, Feedforwar ward Neural Network Neural Network Neural Network Neural Seed (MLPs), Representescent, Feedforwar ward Neural Network Neural Seed (MLPs), Represente Neural Neural Seed (MLPs), Represente Neural Neural Neural Seed (MLPs), Represente Neural Neural Neural Seed (MLPs), Represente Neural Ne	ons, Perceptron Lear tation Power of M rd Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, E noise at input, Ensertter activation function. Processing:	ming LPs, orks, ation GD, Datamble ions,	Hours 5			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descent Stochastic GD, A Regularization: Regularization: Set augmentation methods, Dropos Better weight ini Deep Learning for Eigen values and	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent on (GD), Momentu AdaGrad, RMSProp, Bias Variance Trade on, Parameter sharing out. Greedy Layer w tialization methods, for word encoding- d eigen vectors, B	ing Logic, Perceptro (MLPs), Representescent, Feedforwar ward Neural Network Manager Manager Pre-training, Be Batch Normalization (Natural Language Matural Communication (Matural Language Matural Communication (Matural Communicat	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, Enoise at input, Ensertter activation function. Processing: ponent Analysis and	ming LPs, orks, ation GD, Datamble ions,	5 4			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descen Stochastic GD, A Regularization: Regularization: I set augmentation methods, Dropor Better weight ini Deep Learning in Eigen values an interpretations,	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent on (GD), Momenta AdaGrad, RMSProp, Bias Variance Trade on, Parameter sharing out. Greedy Layer w tialization methods, for word encoding- od eigen vectors, B Singular Value	ing Logic, Perceptro (MLPs), Representations of the secondary of the secon	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, Enoise at input, Ensertter activation function. Processing: ponent Analysis and Learning Vector	ming LPs, orks, ation GD, Datamble ions, d its orial	Hours 5			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descention Stochastic GD, A Regularization: Regularization: Set augmentation methods, Dropool Better weight ini Deep Learning in Eigen values an interpretations, Representations	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent on (GD), Momenta AdaGrad, RMSProp, Bias Variance Trade on, Parameter sharing out. Greedy Layer w tialization methods, for word encoding- d eigen vectors, B Singular Value Of Words: One ho	ing Logic, Perceptro (MLPs), Representations of the secondary of the secon	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, Enoise at input, Ensertter activation function. Processing: ponent Analysis and	ming LPs, orks, ation GD, Datamble ions, d its orial	5 4			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descent Stochastic GD, A Regularization: Regularization: Set augmentation methods, Droport Better weight ini Deep Learning of Eigen values an interpretations, Representations word representations	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent on Gradient	ing Logic, Perceptro (MLPs), Representescent, Feedforwar ward Neural Network Neural Network Neural Network Neural Neural Network Neural	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, Donoise at input, Ensertter activation function. Processing: ponent Analysis and Learning Vectowords, SVD for lear	ming LPs, orks, ation GD, Datamble ions, d its orial	5 4			
I III IV	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descen Stochastic GD, A Regularization: Regularization: Regularization: Set augmentation methods, Dropos Better weight ini Deep Learning of Eigen values an interpretations, Representations word representation	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent on (GD), Momenta AdaGrad, RMSProp, Bias Variance Trade on, Parameter sharing out. Greedy Layer w tialization methods, for word encoding- od eigen vectors, B Singular Value Of Words: One ho ion. Neural Networks fo	ing Logic, Perceptro (MLPs), Representation of variable Contents ing Logic, Perceptro (MLPs), Representation of variable Contents of Matural Network (MLPs), Representation of variable Contents of Matural Language Decomposition, trepresentation of variable Computer Vision:	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, Enoise at input, Ensertter activation function. Processing: ponent Analysis and Learning Vectowords, SVD for lear	ming LPs, orks, ation GD, Datamble ions, d its orial ming	5 4			
I	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descen Stochastic GD, A Regularization: Regularization: Set augmentation methods, Dropor Better weight ini Deep Learning of Eigen values an interpretations, Representations word representat Convolutional N	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent on (GD), Momenta AdaGrad, RMSProp, Bias Variance Trade on, Parameter sharing ont. Greedy Layer we tialization methods, for word encoding- od eigen vectors, B Singular Value Of Words: One ho ion. Neural Networks for eural Networks, Lel	ing Logic, Perceptro (MLPs), Representation Neural Network Neural Network Neural Network Neural Network Neural Neural Network Neural Ne	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, Enoise at input, Ensertter activation function. Processing: ponent Analysis and Learning Vectowords, SVD for lear	ming LPs, orks, ation GD, Datamble ions, d its orial ming	Hours 5			
I III IV	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descent Stochastic GD, A Regularization: Regularization: Set augmentation methods, Dropot Better weight ini Deep Learning for Eigen values an interpretations, Representations word representation Convolutional N ResNet, Visualization	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent on (GD), Momenta AdaGrad, RMSProp, Bias Variance Trade on, Parameter sharing out. Greedy Layer w tialization methods, for word encoding- d eigen vectors, B Singular Value Of Words: One ho ion. Neural Networks for eural Networks, Lel zing Convolutional	ing Logic, Perceptro (MLPs), Representation Neural Networks and tying, Injecting Presentation Natural Language I asis, Principal Computer Vision: Net, AlexNet, ZF-Net Neural Networks, Computer Networks,	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, Enoise at input, Ensertter activation function. Processing: ponent Analysis and Learning Vectowords, SVD for lear et, VGGNet, GoogLe Guided Backpropagar	ming LPs, orks, ation GD, Datamble ions, d its orial ming	5 4 4 4			
I III IV	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descent Stochastic GD, A Regularization: Regularization: Set augmentation methods, Dropot Better weight ini Deep Learning for Eigen values an interpretations, Representations word representation Convolutional N ResNet, Visualization	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent on Gradient	ing Logic, Perceptro (MLPs), Representation Neural Network Neural Network Neural Network Neural Network Neural Neural Network Neural Ne	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, Enoise at input, Ensertter activation function. Processing: ponent Analysis and Learning Vectowords, SVD for lear et, VGGNet, GoogLe Guided Backpropagar	ming LPs, orks, ation GD, Datamble ions, d its orial ming	5 4 4 4			
I III IV	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descer Stochastic GD, A Regularization: Regularization: Is set augmentation methods, Dropor Better weight ini Deep Learning of Eigen values an interpretations, Representations word representat Convolutional N ResNet, Visualiz Deep Dream, De Recurrent Neuro	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient D Power of Feedfor on Gradient Descent on Gradient	ing Logic, Perceptro (MLPs), Representation Neural Networks and Neural Networks, Capable Computer Vision: In Computer Vision: Network AlexNet, ZF-Neural Networks, Capable Computer Neural Ne	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, Donoise at input, Ensertter activation function. Processing: ponent Analysis and Learning Vectowords, SVD for lear set, VGGNet, GoogLe Guided Backpropagasetworks	ming LPs, orks, ation GD, Datamble ions, d its orial ming	5 4 4 4			
I II III V	McCulloch Pitts Algorithm, Mul Sigmoid Neuro Representation algorithm. Optimizations in Gradient Descent Stochastic GD, A Regularization: Regularization: Regularization: Set augmentation methods, Dropot Better weight ini Deep Learning in Eigen values an interpretations, Representations word representat Convolutional N Convolutional N ResNet, Visualiz Deep Dream, De Recurrent Neuro	Mod of Neural Networks Neuron, Threshold tilayer Perceptrons ons, Gradient De Power of Feedfor on Gradient Descent on Gradient	ing Logic, Perceptro (MLPs), Representation Neural Networks and Neural Networks, Capable Computer Vision: In Computer Vision: Network AlexNet, ZF-Neural Networks, Capable Computer Neural Ne	ons, Perceptron Lear tation Power of Mord Neural Netwoworks. Backpropagasterov Accelerated on in Adam. on, Early stopping, Enoise at input, Ensertter activation function. Processing: ponent Analysis and Learning Vectowords, SVD for lear et, VGGNet, GoogLe Guided Backpropagasterworks time(BPTT), Vanish	ming LPs, orks, ation GD, Datamble ions, d its orial ming	5 4 4 4			

	Concepts, Tools and Techniques to Build Intelligent Systems", 2 nd Edition, O'Reilly,2019						
2	Eugene Charniak, "Introduction to Deep Learning, The MIT Press Cambridge", 1st Edition, 2019						
	References						
1	Ian Goodfellow, Yoshua Bengio and Aoron Courville "Deep Learning", The MIT Press						
1	Cambridge, Massachusetts London, England, 2017						
Useful Links							
1	All Modules taken from below link						
1	https://www.classcentral.com/course/swayam-deep-learning-iitropar-43579						

CO-PO Mapping							
		Programme Outcomes (PO)					
	PO1	PO2	PO3	PO4	PO5	PO6	
CO1		1			1		
CO2		2		2			
CO3	2		1				

Assessment Plan based on Bloom's Taxonomy Level							
Bloom's Taxonomy Level	T1	T2	ESE	Total			
Remember	To be used minimum						
Understand	To be used minimum						
Apply	10	10	20	40			
Analyze	5	5	15	25			
Evaluate	5	5	15	25			
Create			10	10			
Total	20	20	60	100			

Walchand College of Engineering, Sangli							
(Government Aided Autonomous Institute)							
AY 2021-22							
	Course Information						
Programme	M. Tech. (CS and IT)						
Class, Semester	First Year M. Tech., Sem II						
Course Code	5IT575						
Course Name	Activity Based Lab for Big Data Computing						
Desired Requisites:							

Teaching Scheme		Examination Scheme (Marks)					
Lecture	-	LA1	LA2	ESE	Total		
Tutorial	-	30	30	40	100		
Practical	2 Hrs/Week						
Interaction	-		Credits: 1				

	Course Objectives						
1	To demonstrate the big data computing using Apache Hadoop						
2	To experiment the distributed file system and its interfacing						
3	3 To solve real world challenges using big data analytics						
	Course Outcomes (CO) with Bloom's Taxonomy Level						
At the	end of the course, the students will be able to,						
CO1	CO1 Apply the concepts of big data computing for data analytics Apply						
CO2	Identify the characteristics of datasets in big data	Apply					
CO3	Evaluate scaling techniques to compute the big data	Evaluate					

List of Experiments / Lab Activities

List of Experiments:

Activities are to be carried out individually.

Each student will perform the activity based on course on following areas.

- 1. Implement the following file management tasks in Hadoop: Adding Files and Directories, Retrieving Files, Deleting Files
- 2. Exploring various shell commands in Hadoop.
- 3. Industry Problem Statement(if any)
- 4. To implement basic Word Count Map-Reduce program to understand Map Reduce Paradigm with number of occurrences of each word appearing in an input file and perform a MapReduce Job for word search count (look for specific keywords in a file).
- 5. Implement Map Reduce program that mines weather data (or any real-time data set). Weather sensors collecting data every hour at many locations across the globe gather large volume of log data, which is a good candidate for analysis with MapReduce, since it is semi structured and record-oriented

Student should perform the activities on the basis of the real-time applications in the subjects and submit the work with code, PPT, PDF, Text report document & reference material or on online GitHub. Students should maintain activity log book containing weekly progress.

	Text Books				
1	Prajapati Vignesh, "Big Data Analytics with R and Hadoop", Packt Publishing, 1st Edition, 2013				
2	Minelli Michael, Chambers Michehe, "Big Data, Big Analytics: Emerging Business Intelligence and Analytic Trends for Today's Business", Ambiga Dhiraj, Wiely CIO Series, 1 st Edition, 2013				
	and Analytic Trends for Today's Business', Amolga Diniaj, wiery Clo Series, 1' Edition, 2013				
	References				
1	Franks Bill, "Taming the Big Data Tidal Wave: Finding Opportunities in Huge Data Streams				
1	with Advanced Analytics", Wiley and SAS Business Series, 1st Edition, 2012				
	Useful Links				
1	Module I, II, III, IV, V, VI				
1	https://nptel.ac.in/courses/106/104/106104189/				

CO-PO Mapping							
		Programme Outcomes (PO)					
	PO1 PO2 PO3 PO4 PO5 PO6						
CO1	2		1				
CO2	3 2 2						
CO3	1		1				

There are three components of lab assessment, LA1, LA2 and Lab ESE.

IMP: Lab ESE is a separate head of passing. LA1, LA2 together is treated as In-Semester Evaluation.

Assessment	Based on	Conducted by	Typical Schedule (for 26-week Sem)	Marks		
LA1	Lab activities,	Lab Course	During Week 1 to Week 6	30		
LAI	attendance, journal	Faculty	Marks Submission at the end of Week 6	30		
LA2	Lab activities,	Lab Course	During Week 7 to Week 12	30		
LAZ	attendance, journal	Faculty	Marks Submission at the end of Week 12	30		
Lab ECE	Lab activities,	Lab Course	During Week 15 to Week 18	40		
Lab ESE	attendance, journal	Faculty	Marks Submission at the end of Week 18	40		

Week 1 indicates starting week of a semester. The typical schedule of lab assessments is shown, considering a 26-week semester. The actual schedule shall be as per academic calendar. Lab activities/Lab performance shall include performing experiments, mini-project, presentations, drawings, programming and other suitable activities, as per the nature and requirement of the lab course. The experimental lab shall have typically 8-10 experiments.

Assessment Plan based on Bloom's Taxonomy level							
Bloom's Taxonomy Level	LA1	LA2	Lab ESE	Total			
Remember	To be used minimum						
Understand	To be used minimum						
Apply	10	10	10	30			
Analyze	10	10	10	30			
Evaluate	5	5	10	20			
Create	5	5	10	20			
Total	30	30	40	100			

	Walchand College of Engineering, Sangli				
	(Government Aided Autonomous Institute)				
	AY 2020-21				
	Course Information				
Programme M. Tech. (CS and IT)					
Class, Semester First Year M. Tech., Sem II					
Course Code	Course Code 5IT576				
Course Name	Course Name Activity Based Lab for High Performance Computing				
Desired Requisites:	Computer Algorithm				

Teaching Scheme		Examination Scheme (Marks)				
Lecture -		LA1	LA2	ESE	Total	
Tutorial	-	30	30	40	100	
Practical	2 Hrs/Week					
Interaction	-		Cred	its: 1		

	Course Objectives					
1	To elaborate the concepts of process and thread in high performance computing					
2	To evaluate the performance of parallel programs with sequential program					
3	To compare multi-core and many-core architectures					
	Course Outcomes (CO) with Bloom's Taxonomy Level					
At the	At the end of the course, the students will be able to,					
CO1	Apply the parallel algorithm to solve the problem	Apply				
CO2	Implement the parallel algorithms for performance parameters	Apply				
CO3	Develop the appropriate parallel algorithm to speed up the computation	Create				

List of Experiments / Lab Activities

Lab Activities:

Activities are to be carried out individually.

Each student will perform the activity based on course on following areas.

Implementations are expected using OpenACC platform

- 1. Implement PI Calculation.
- 2. Implement Matrix Transpose Program.
- 3. Write a program to find the factorial of a given number.
- 4. Write a program to find squares of array elements.
- 5. Implement odd-even Sort.
- 6. Implement Quick Sort.
- 7. Program on vector computation.
- 8. Study of Profiling tools.

Stude	ent should perform the activities on the basis of the real-time applications in the subjects and submit								
the w	the work with code, PPT, PDF, Text report document & reference material or on online GitHub.								
1	ents should maintain activity log book containing weekly progress.								
	J. J								
	Text Books								
1	Anath Grama, Ansul Gupta, George Karypis, Vipin Kumar, "Introduction to parallel computing", Second Edition, Pearson Education								
	References								
1	Horrowitz, SahniRajasekaran, "Computer Algorithms", Computer Science, W. H. Freeman and company Press, New York,								
	Useful Links								
	https://www.coursera.org/learn/parprog1?ranMID=40328&ranEAID=*GqSdLGGurk&ranSiteID								
1	=.GqSdLGGurk-ntwHfWI_xX32aIgZXdr9Ug&siteID=.GqSdLGGurk-								
1	ntwHfWI_xX32aIgZXdr9Ug&utm_content=10&utm_medium=partners&utm_source=linkshare&utm_campaign=*GqSdLGGurk#syllabus								

CO-PO Mapping							
		Programme Outcomes (PO)					
	PO1 PO2 PO3 PO4 PO5 PO6						
CO1			3				
CO2	2	2		1			
CO3		1				2	

There are three components of lab assessment, LA1, LA2 and Lab ESE.

IMP: Lab ESE is a separate head of passing. LA1, LA2 together is treated as In-Semester Evaluation.

Assessment	Based on	Conducted by	Typical Schedule (for 26-week Sem)	Marks
LA1	Lab activities,	Lab Course	During Week 1 to Week 6	30
LAI	attendance, journal	Faculty	Marks Submission at the end of Week 6	30
LA2	Lab activities,	Lab Course	During Week 7 to Week 12	30
LAZ	attendance, journal	Faculty	Marks Submission at the end of Week 12	30
Lab ECE	Lab activities,	Lab Course	During Week 15 to Week 18	40
Lab ESE	attendance, journal	Faculty	Marks Submission at the end of Week 18	40

Week 1 indicates starting week of a semester. The typical schedule of lab assessments is shown, considering a 26-week semester. The actual schedule shall be as per academic calendar. Lab activities/Lab performance shall include performing experiments, mini-project, presentations, drawings, programming and other suitable activities, as per the nature and requirement of the lab course. The experimental lab shall have typically 8-10 experiments.

Assessment Plan based on Bloom's Taxonomy level							
Bloom's Taxonomy Level	LA1	LA2	Lab ESE	Total			
Remember	To be used minimum						
Understand	To be used minimum						
Apply	10	10	10	30			
Analyze	10	10	10	30			
Evaluate	5	5	10	20			
Create	5	5	10	20			
Total	30	30	40	100			

	Walchand College of Engineering, Sangli					
(Government Aided Autonomous Institute)						
	AY 2021-22					
	Course Information					
Programme	M. Tech. (CS and IT)					
Class, Semester	First Year M. Tech., Sem II					
Course Code	5IT577					
Course Name	Activity Based Lab for Deep Learning					
Desired Requisites:						

Teachin	g Scheme	Examination Scheme (Marks)					
Lecture	-	LA1 LA2 ESE Total					
Tutorial	-	30	30	40	100		
Practical	2 Hrs/Week						
Interaction	-	Credits: 1					

	Course Objectives						
1	To elaborate the models of Deep Learning						
2	To compare the applications of Deep Learning with performance parameters						
3	3 To interpret the problem to solve using Deep Learning						
	Course Outcomes (CO) with Bloom's Taxonomy Level						
At the	end of the course, the students will be able to,						
CO1	Apply the Deep Learning model for suitable applications	Apply					
CO2	Implement deep neural network systems	Apply					
CO3	Build Deep Learning model for solving real world application	Evaluate					

List of Experiments / Lab Activities

List of Experiments:

Activities are to be carried out individually.

Each student will perform the activity based on course on following areas.

- 1. Perform the perceptron learning algorithm
- 2. Perform the gradient descent algorithm and its types
- 3. Perform the feedforward neural networks
- 4. Perform the AdaGrad algorithm
- 5. Perform L2 regularization and ensemble methods

https://nptel.ac.in/courses/106/104/106104189/

- 6. Study and explain the better activation functions and better weight initialization methods
- 7. Perform principal component analysis and its interpretation
- 8. Perform bag of words and skip gram model
- 9. Perform CNN related algorithms (LeNet, AlexNet, ZF-Net, VGGNet, GoogleNet, ResNet, etc...)
- 10. Perform object detection using CNN
- 11. Perform YOLO algorithm
- 12. Perform RNN algorithm
- 13. State and explain with example Back propagation through time (BPTT)

1.	13. State and explain with example back propagation unough time (b) 11)						
	Text Books						
1	Prajapati Vignesh, "Big Data Analytics with R and Hadoop", Packt Publishing, 1st Edition, 2013						
2	Minelli Michael, Chambers Michehe, "Big Data, Big Analytics: Emerging Business Intelligence						
	and Analytic Trends for Today's Business", Ambiga Dhiraj, Wiely CIO Series, 1st Edition, 2013						
	References						
1	Franks Bill, "Taming the Big Data Tidal Wave: Finding Opportunities in Huge Data Streams						
1	with Advanced Analytics", Wiley and SAS Business Series,1st Edition, 2012						
	Useful Links						
1	Module I, II, III, IV, V, VI						

CO-PO Mapping							
		Programme Outcomes (PO)					
	PO1 PO2 PO3 PO4 PO5 PO6						
CO1	2		2		2		
CO2		3		3			
CO3	1		1				

There are three components of lab assessment, LA1, LA2 and Lab ESE.

IMP: Lab ESE is a separate head of passing. LA1, LA2 together is treated as In-Semester Evaluation.

Assessment	Based on	Conducted by	Typical Schedule (for 26-week Sem)	Marks
LA1	Lab activities,	Lab Course	During Week 1 to Week 6	30
LAI	attendance, journal	Faculty	Marks Submission at the end of Week 6	30
LA2	Lab activities,	Lab Course	During Week 7 to Week 12	30
LAZ	attendance, journal	Faculty	Marks Submission at the end of Week 12	30
Lab ESE	Lab activities,	Lab Course	During Week 15 to Week 18	40
Lauese	attendance, journal	Faculty	Marks Submission at the end of Week 18	40

Week 1 indicates starting week of a semester. The typical schedule of lab assessments is shown, considering a 26-week semester. The actual schedule shall be as per academic calendar. Lab activities/Lab performance shall include performing experiments, mini-project, presentations, drawings, programming and other suitable activities, as per the nature and requirement of the lab course. The experimental lab shall have typically 8-10 experiments.

Assessment Plan based on Bloom's Taxonomy level							
Bloom's Taxonomy Level	LA1	LA2	Lab ESE	Total			
Remember	To be used minimum						
Understand	To be used minimum						
Apply	10	10	10	30			
Analyze	10	10	10	30			
Evaluate	5	5	10	20			
Create	5	5	10	20			
Total	30	30	40	100			

				ege of Engineering ided Autonomous In					
			,	AY 2021-22	,				
	Course Information Programme M. Tech. (CS and IT)								
Progra									
Class,	Semes								
Course Code									
Course	Nam	e	Open Elective - 4:	Machine Learning	and Applications				
Desire	Desired Requisites:								
Tea	aching	Scheme		Examination S	cheme (Marks)				
Lectur		2 Hrs/week	T1	T2	ESE	Total			
Tutoria		-	20	20	60	100			
Practic	cal	-		1	'	ı			
Interac		-		Cred	lits: 2				
			I						
			Cou	ırse Objectives					
1	To ex	plain the conc		unsupervised machin	ne learning techniqu	es.			
2			s machine learning			- 144 -			
3			<u>_</u>	using appropriate m	nachine learning tec	hniques			
3	10 01			O) with Bloom's Ta		imques.			
At the	end of		e students will be ab		Automy Level				
CO1	Sumi			ng algorithms for	Regression and	Understand			
CO2			aamina alaamithm f	an mantiaulan muahlan	20	A mm1r:			
CO2				or particular problem		Apply			
CO3	Struc	turing Machine	e Learning algorithi	ns with performance	e parameters.	Analyze			
36 11			3.6.1.1	Q 4 4		TT			
Modul		4 3 4	Module	e Contents		Hours			
I	P	•	•	•	Theory, Linear Algebra, Convex Optimization, Statistical				
	Decision Theory - Regression, Classification, Bias Variance trade off.								
	– I K	egression:		Sirication, Dias vari	ance trade on.				
II	L:	lethods, Princip	on, Multivariate Re pal Component Re	egression, Subset Se gression, Partial Lea	lection, Shrinkage ast squares, Linear	4			
III	L: M: C: A In	inear Regressic lethods, Princi lassification, L rtificial Neura troduction, Ea	on, Multivariate Repal Component Repositic Regression, al Networks: rly Models, Percept	egression, Subset Se gression, Partial Lea Linear Discriminant ron Learning, Backp	lection, Shrinkage ast squares, Linear tAnalysis.	5			
	L. M. C. C. A. In In In C.	inear Regression (lethods, Principles lassification, Letroduction, Eatitalization, Transcriptor (legorithms: lecision Trees, unctions, Categorithms).	on, Multivariate Repal Component Repogistic Regression, al Networks: rly Models, Perceptraining & Validation Regression Trees, Sorical Attributes, M	egression, Subset Se gression, Partial Lea Linear Discriminant ron Learning, Backp	lection, Shrinkage ast squares, Linear Analysis. propagation, Pruning loss ing	•			
III	L. M. C. A. In	inear Regression lethods, Principles lassification, Letroduction, Easterduction, Transcription Trees, anctions, Categoralues, Decision Lachines, earning Theorootstrapping &	on, Multivariate Repal Component Regogistic Regression, al Networks: rly Models, Percept aining & Validation Regression Trees, Socical Attributes, Marrees – Instability ry: c Cross Validation, esemble Methods - I	egression, Subset Segression, Partial Lea Linear Discriminant ron Learning, Backp h. Stopping Criterion & Jultiway Splits, Miss	lection, Shrinkage ast squares, Linear Analysis. propagation, Pruning loss ing s, Support Vector easures, ROC	5			
III	Land Market Mark	inear Regression lethods, Principles lassification, Letrificial Neural troduction, Eastitialization, Transcriptions (actions, Categoral alues, Decision Lachines, Learning Theorem ootstrapping & Learning, Boostilustering: Learning Clustering (actional Clustering)	on, Multivariate Repal Component Regogistic Regression, al Networks: rly Models, Percept aining & Validation Regression Trees, Sorical Attributes, Ma Trees – Instability ry: c Cross Validation, of the semble Methods - Instability	egression, Subset Segression, Partial Lea Linear Discriminant ron Learning, Backpa. Stopping Criterion & Jultiway Splits, Miss Evaluation Measure Class Evaluation Me Bagging, Committee Clustering, Birch Al	lection, Shrinkage ast squares, Linear tAnalysis. Propagation, Pruning loss ing s, Support Vector easures, ROC Machines and	5			
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III IV V VI 1	L. M. C. C. A. In	inear Regression lethods, Principlessification, Letrificial Neural troduction, Eastitialization, Trogorithms: ecision Trees, anctions, Categalues, Decision lachines, earning Theorem ootstrapping & arve, MDL, Entacking, Boosti lustering: eartitional Clusting or Hastie, Robert Park (1997) and Edition of Hastie Park (1997) and Edition of Hastie Park (1997) and Edition of Hastie Park (1997) and Edition of Has	on, Multivariate Repal Component Regogistic Regression, al Networks: rly Models, Percept raining & Validation Regression Trees, Sorical Attributes, Ma Trees – Instability ry: c Cross Validation, osemble Methods - Fing ering, Hierarchical sity-based Clusterin ert Tibshirani, Jeroon, 2009.	egression, Subset Segression, Partial Lea Linear Discriminant ron Learning, Backpa. Stopping Criterion & Jultiway Splits, Miss Evaluation Measure Class Evaluation Measure Class Evaluation Measure Clustering, Birch Alg Text Books me H. Friedman, " References	lection, Shrinkage ast squares, Linear tAnalysis. Propagation, Pruning loss ing s, Support Vector easures, ROC Machines and gorithm, CURE	5 5 4 4 atistical Learning"			
III IV V VI	L. M. C. C. A. In	inear Regression lethods, Principlessification, Letrificial Neural troduction, Eastitialization, Trogorithms: ecision Trees, anctions, Categalues, Decision lachines, earning Theorem ootstrapping & arve, MDL, Entacking, Boosti lustering: eartitional Clusting or Hastie, Robert Park (1997) and Edition of Hastie Park (1997) and Edition of Hastie Park (1997) and Edition of Hastie Park (1997) and Edition of Has	on, Multivariate Repal Component Regogistic Regression, al Networks: rly Models, Percept raining & Validation Regression Trees, Sorical Attributes, Ma Trees – Instability ry: c Cross Validation, osemble Methods - Fing ering, Hierarchical sity-based Clusterin ert Tibshirani, Jeroon, 2009.	egression, Subset Segression, Partial Lea Linear Discriminant ron Learning, Backpa. Stopping Criterion & Sultiway Splits, Miss Evaluation Measure Class Evaluation Measure Class Evaluation Measure Clustering, Birch Alg Text Books me H. Friedman, ""	lection, Shrinkage ast squares, Linear tAnalysis. Propagation, Pruning loss ing s, Support Vector easures, ROC Machines and gorithm, CURE	5 4 4 attistical Learning"			
III IV V VI 1	L. M. C. C. A. In	inear Regression lethods, Principlassification, Letrificial Neural troduction, Eastitialization, Trogorithms: ecision Trees, anctions, Categalues, Decision lachines, earning Theolootstrapping & arve, MDL, Entacking, Boostitustering: artitional Clustilgorithm, Densor Hastie, Robleger, 2nd Edition letopher Bishop, attopher Bishop,	on, Multivariate Repal Component Regogistic Regression, al Networks: rly Models, Percept raining & Validation Regression Trees, Sorical Attributes, Ma Trees – Instability ry: c Cross Validation, assemble Methods - Hing ering, Hierarchical sity-based Clusterin ert Tibshirani, Jeroon, 2009.	egression, Subset Segression, Partial Lea Linear Discriminant ron Learning, Backpa. Stopping Criterion & Jultiway Splits, Miss Evaluation Measure Class Evaluation Measure Class Evaluation Measure Clustering, Birch Alg Text Books me H. Friedman, " References	lection, Shrinkage ast squares, Linear tAnalysis. propagation, Pruning loss ing s, Support Vector easures, ROC Machines and gorithm, CURE The Elements of Statements of Statements and Statements of Statements	5 5 4 4 Autistical Learning			

2	https://web.stanford.edu/~hastie/Papers/ESLII.pdf
3	http://users.isr.ist.utl.pt/~wurmd/Livros/school/Bishop%20- %20Pattern%20Recognition%20And%20Machine%20Learning%20- %20Springer%20%202006.pdf

CO-PO Mapping							
	Programme Outcomes (PO)						
	PO1 PO2 PO3 PO4 PO5 PO6						
CO1			1				
CO2	2	1		2	2		
CO3	3		2				

Assessment Plan based on Bloom's Taxonomy level								
Bloom's Taxonomy Level	LA1	LA2	Lab ESE	Total				
Remember	To be used minimum							
Understand	To be used minimum							
Apply	10	10	10	30				
Analyze	10	10	10	30				
Evaluate	5	5	10	20				
Create	5	5	10	20				
Total	30	30	40	100				