

SAGI RAMA KRISHNAM RAJU ENGINEERING COLLEGE (AUTONOMOUS)

(Affiliated to JNTUK, Kakinada), (Recognized by AICTE, New Delhi) UG Programmes CE,CSE,ECE,EEE,IT & ME are Accredited by NBA, Accredited by NAAC with A⁺ CHINNA AMIRAM (P.O):: BHIMAVARAM :: W.G.Dt., A.P., INDIA :: PIN: 534 204

Estd:1980

Regula	II	III / IV - B.Tech. I - Semester							
	COMPUTER SCIENCE AND DESIGN								
	SCHEME OF INSTRUCTION & EXAMINATION (With effect from 2021-22 admitted Batch onwards)								
Course Code	Course Name	Catego ry	Cr	L	Т	Р	Int. Marks	Ext. Marks	Total Marks
B20CS3101	Data Warehousing and Data Mining	PC	3	3	0	0	30	70	100
B20CD3101	Computer Aided Design	PC	3	3	0	0	30	70	100
B20AM3103	Machine Learning	PC	3	3	0	0	30	70	100
#PE-I	Professional Elective -I	PE	3	3	0	0	30	70	100
#OE-I	Open Elective-I	OE	3	3	0	0	30	70	100
B20CS3108	Da <mark>ta Mining</mark> Lab	PC	1.5	0	0	3	15	35	50
B20CD3105	Advance Data Structures La	b PC	1.5	0	0	3	15	35	50
B20HS3102	Soft Skills (Skill Oriented Course)	SOC	2	١Ġ	0	2	EGE	50	50
B20MC3105	Digital Marketing Lab	MC	0	0	0	3			
B20CD3106	Summer Internship	PR	1.5					50	50
	ΤΟΤΑ					8	180	520	700

	Course Code	Course				
	B20CS3104	Artificial Intelligence				
#DE I	B20CD3102	Object Oriented Analysis and Design				
#PE-I	B20CD3103	Human Computer Interaction				
	B20CD3104	Visual Design and Communication				
	B20AM3104	Internet of Things				
#OF I	Student has to s	tudy one Open Elective offered by CE or ECE or EEE or ME or				
#OE-I	S&H from the list enclosed.					

Co	ode	Category	L	Т	Р	С	I.M	E.M	Exam	
B20C	S3101	PC	3			3	30	70	3 Hrs.	
		1			1	I		1		
	DATA WAREHOUSING AND DATA MINING									
	(Common to CSE & CSD)									
Cours	Course Objectives: Students are expected to									
1	The d	ata warehouse	concept	s, architec	ture and d	lata mining	g techniqu	es.		
2	The types of data, their characteristics, cleaning and transformation of data for analysis.									
2	The principles of statistics, information theory, databases, machine learning and other areas for									
3	design	n and impleme	ntation o	of data min	ning techn	iques.				
4	Assoc	iation rule min	ning, cla	ssification	and clust	ering tech	niques.			
Course	e Outco	mes: At the en	nd of the	course st	udents wil	l be able t	0			
S No				0	taama				Knowledge	
3. NU				U	ncome				Level	
1.	Identi	fy the importa	nce of d	ata mining	g and able	to interpre	et the data.		K3	
2	Expla	in the concept	s of dat	a warehou	ising & O	LAP tech	nology and	d apply data	К3	
	pre pr	ocessing techr	niques.							
3	Form	ulate and appl	y associ	ation rule	mining al	gorithms	and their J	performance	K3	
	evalua	ation metrics o	n sampl	e datasets.		1.4	6	- 1 -		
4	Form	late and apply	y classif	ication alg	gorithms a	nd their p	erformanc	e evaluation	K3	
	Apply	and compar	alasels.	oning hie	erarchical	density 1	pased and	grid based		
5	cluste	ring algorithm	s.	oning, inc	crarentear,	density	Jused and	gild based	K4	
		00								
				S	SYLLABU	U S				
	I	ntroduction to) Data N	Iining: W	hat is Dat	a Mining?	Importan	ce of Data N	lining, kinds of	
UNIT	'-I D	ata and Patter	ns to be	Mined, Da	ata Mining	g Applicati	ions, Majo	or Issues in I	Data Mining.	
(10 H	rs) G	etting to Kno	ow You	r Data: D	ata Objec	ts and Att	ribute Typ	pes, Statistic	al Descriptions	
	0	of Data, Estimating Data Similarity and Dissimilarity.								
	D	ata Preproc	essing:	Data (Cleaning,	Data Ir	ntegration,	Data Re	duction, Data	
UNIT	т т	ransformation	and Dat	a Discreti	zation.					
(10 H	$\frac{\mathbf{n}}{\mathbf{rs}}$ D	ata Warehou	sing and	l Online A	Analytical	Processi	ng:			
	••• B	asic Concepts	s of Dat	ta wareho	ouse, Data	Warehou	ise Model	ling using l	Data Cube and	
	OLAP, Data warehouse Design and Usage, Data warehouse Implementation.							l.		
	N	lining Freque	ent Patt	erns, Ass	sociations	and Cor	relations:	Basic Con	cepts, Frequent	
UNIT-	III It	emset Mining	Method	s: Apriori	Algorith	n, Associa	ation Rule	Generation	Improvements	
(10 H	\mathbf{rs} to	Apriori, FP-	Growth	Approach	i, Mining	Frequent	Itemsets U	sing Vertica	al Data Format,	
	Closed and Max Patterns, Pattern Evaluation Methods: Mining in multilevel,							utilevel, multi-		

		dimensional space.							
UNI	Γ-IV	Classification: Basic Concepts, Decision Tree Induction, Bayes Classification, Rule-							
(10 F	Hrs)	Based Classification, Model Evaluation and Selection, Techniques to Improve							
(101		Classification Accuracy, Advanced Methods: Classification by Back Propagation.							
		Cluster Analysis: Basic Concepts, Partitioning Methods: K-Means, K-Medoids,							
UNI	Γ-V	Hierarchical Methods: Agglomerative versus Divisive Hierarchical Clustering, Distance							
(12 H	Hrs)	Measures in Algorithmic Methods, BIRCH, Density Based Methods: DBSCAN, OPTICS,							
		Grid Based Methods: STING.							
Textl	book:								
1	Data	Mining Concepts and Techniques by Jiawei Han, Micheline Kamber and Jian Pei –Morgan							
1.	Kau	fmann publishers –3 rd Edition,2012.							
Refer	rence	Books:							
1.	Intro	oduction to Data Mining, Adriaan, Addison Wesley Publication, 2016							
ر د	Data	Mining Techniques, A.K.Pujari, University Press Data mining concepts by Tan, Steinbech,							
۷.	and	Vipin Kumar - Pearson Edu publishers,2001.							
3	Data	Mining –Introductory and Advanced by Margarett Dunham Pearson Education							
5.	publ	ishers,2006.							
Δ	Pang	g-Ning Tan, Michael Steinbach and Vipin Kumar, Introduction to Data Mining,							
т.	Pear	son,2016.							
		ENGINEERING COLLEGE							
		Estd. 1980 AUTONOMOUS							

(Code	Category	L	Т	Р	С	I.M	E.M	Exam	
B20CD3101		PC	3			3	30	70	3 Hrs.	
	COMPUTER AIDED DESIGN									
	(For CSD)									
Cour	Course Objectives: The objective of the course is to enable students to									
1.	1. Provide basic foundation in computer aided design									
2.	Understand the fundamentals used to create and manipulate geometric models									
3.	3. Get acquainted with the basic CAD software designed for geometric modeling									
	-				0	0		0		
Cour	rse Outc	omes: At the end	d of the co	irse stud	lents will	be able to)			
S.							-		Knowledge	
No				Outco	me				Level	
1.	Apply	he fundamentals	s of CAD a	nd Com	puter Sv	stems			К3	
2.	Unders	tand CAD syster	n hardwar	e and sot	ftware.				К2	
3.	Apply	geometric model	ling techni	aues for	complex	part and	shape de	sign.	К3	
4.	Apply	2 D & 3D transfo	rmations f	or part a	nd shape	alteration	<u>s.</u>	0	K3	
5.	Apply	he applications	of Artificia	1 Intellig	ence and	l Expert S	vstems in		K3	
	1 pp y				,once une	Enperes	JScenis n			
				SVI	LLABUS					
	1	Fundamental of	CAD	511		,				
		ntroduction. Th	ne CAD	system.	Reasons	for imp	lementin	g CAD. D	esign Process.	
UN	IT-I	Applications of Computers in Design, Benefits of Computer Aided Design								
(10)	Hrs)	Computer Systems								
]	Introduction, Hardware Components, Mass Storage Devices, Input/output Devices.								
	•	CAD System Ha	ardware							
]	ntroduction, Th	e Design V	Vorkstat	ion, Graj	ohics Terr	ninal, Gr	aphics Term	ninal for CAD,	
UN	п.п (CAD System Co	nfiguration	1						
(10)	Hrs)	CAD System So	ftware							
(10]	ntroduction, Fu	nctions of	a Graph	nics Pack	age, Oper	rating Sy	stem, Graph	nics Standards,	
		Standards for Gr	aphics Pro	grammi	ng, Stand	lard for th	e Exchar	nge of Produ	et Model Data	
	(STEP), Drawing	g Exchange	e Format	t (DXF)					
		Geometric Mod	elling				D			
		ntroduction, Ir	nportance	of Ge	eometric	Modellii	ng, Req	urrements	of Geometric	
UNI	T-III	Modelling, Type	s of Model	S T	Б. <i>с.</i> с.	D	, , .		• •, ,•	
(10	Hrs)	Wire Frame Mi	odelling: v	vire Fran	ne Entiti	es, Repres	entation	of Curves, L	limitations.	
		Solid Modelli	ing: introd	uction, S		nuties, Su	mace Ke	presentation	s, Limitations.	
		Solia Modelling	3: Introduc	$x_0 = \sum_{i=1}^{n} \sum_{j=1}^{n} \sum_{j=1}^{n} \sum_{i=1}^{n} \sum_{j=1}^{n} \sum_{i=1}^{n} \sum_{j=1}^{n} \sum_{i=1}^{n} \sum_{j=1}^{n} \sum_{i=1}^{n} \sum_{j=1}^{n} \sum_{i=1}^{n} \sum_{j=1}^{n} \sum_{j=1}^{n} \sum_{i=1}^{n} \sum_{i=1}^{n$	ona Entit	ies, Solid	wodelli	ng Techniq	ues, Boundary	
		cepresentation, C		e 2011a	Geometr	у.				

		Transformation Systems								
UNI	IT-IV Introduction, Transformation Principles, Two-Dimensional Geometric Transformations:									
(10	Hrs)	Scaling, Translation, Rotation; Three-Dimensional Geometric Transformations: Scaling,								
	Translation, Rotation, Windowing and Clipping									
		Artificial Intelligence and Expert System								
UN	IT-V	Introduction, Artificial Intelligence in CAD, Applications of AI in Design, Expert								
(10	Hrs)	Systems, Structure of an Expert System, Development of an Expert System, Knowledge								
(10	II (5)	Representation, Inference Engine, Characteristics of an Expert System, Example of								
		Expert Systems, Benefits of Expert System								
Text	books:									
1.	CAD	/CAM- Computer Aided Design & Manufacturing, by M.D.Groover & E.W.Zimmer.								
2.	CAD	/ CAM Theory and Practice / Ibrahim Zeid / TMH								
Refe	rence l	Books:								
1.	CAD	/CAM by P.N. Rao/TMH.								
2	Comp	outer Aided Design and Manufacturing by K. Lalit Narayan, K. Mallikarjuna Rao,								
Ζ.	M.M.	M.Sarcar.								
3.	CAD	/CAM/CIM by Radhakrishna, New age international								
e-Re	source									
1.	https:	//nptel.ac.in/courses/112102101								
2.	https:	//nptel.ac.in/courses/112104031								
3.	https:	//onlinecourses.swayam2.ac.in/nou20_cs15/preview_OUS								

Co	ode	Category	CodeCategoryLTPCI.ME.MEx							
B20Al	M3103	PC	3			3	30	70	3 Hrs.	
				MACH	IINE LE	ARNING				
	(Common to AIML & CSD)									
Cours	Course Objectives: Students are expected to									
1	1 Explain the basic concepts and techniques of Machine Learning									
2	Demonstrate regression, classification methods.									
3	Illustra	Illustrate the concepts of dimensionality reduction, artificial neural networks and reinforcement								
	learnin	g								
4	Show	the applicatio	on of mac	hine learn	ing mode	l evaluatio	on and opti	mization tec	hniques	
~										
Course	e Outcor	mes: At the e	nd of the	course stu	idents wil	l be able t	.0			
S. No				Ou	tcome				Knowledge	
1	F 1 '	.1	<u> </u>	1	1 1' '	· .	1 . 1	•	Level	
1	Explai	n the concept	s of ingre	dients and	d prelimi	haries of n	hachine lea	irning	<u>K2</u>	
2	Apply	tree models,	linear mo	dels and o	distance-t	ased mod	els		K3	
3	Demoi	y and constru	ict reature	es and ens	emble mo	odels	n tashnis	una madal	K3	
4	Demo	tion and solo	tion tach	of diffie	insionality	reductio	on techniq	ues, model	K3	
5		the concepts	of artifici	ial neural	networks	reinforce	ment learn	ing	K3	
5	Арріу	the concepts			lictworks	, iciliioice		iiiig	KJ	
			4	ENG	SYLLAB	US	COLL	EGE		
		The ingredi	ents of	machine	learning	: Basic c	oncepts, d	esigning a	learning system,	
		Issues in ma	chine lear	rning, Ty	pes of ma	chine lear	rning, Tas	ks: the prob	elems that can be	
		solved with machine learning, Models: the output of machine learning, Features, the								
UNI	T-I	workhorses of machine learning.								
(12 H	Hrs)	Preliminaries: The curse of dimensionality, Overfitting, Training, Test and Validation								
		sets, The confusion matrix, The accuracy metrics: Accuracy, sensitivity, specificity,								
		precision, red	call, F1 r	neasure, I	ROC curv	ve, Unbala	inced data	sets, Naïve	Bayes Classifier,	
		Some basic s	statistics:	variance	, covarian	ce, bias-v	ariance tra	deoff.		
		T N (
		Tree Models	5: Dec1s10	n Trees.		(N T)	11 1	· c·		
UNI	T-II	Distance Bas	sea Mioa Jola, Th	els: Introd	uction, N	earest Nei	Ignbors cla	ssification.	unation Lociatio	
(10 H	Hrs)	Degression	Support	Vector N	Jachines II	Linear	SVM Cla	sification	Nonlinear SVM	
		Regression, Support Vector Machines: Linear SVM Classification, Nonlinear SVM								
		Features: K	inds of t	feature. F	eature tr	ansformat	ions: Thre	sholding ar	d discretization	
UNIT	Г- III	Normalizatio	n, Incom	plete Feat	ures, Feat	ure consti	uction and	l selection.		
(10 H	Hrs)	Model ense	mbles: V	oting Cl	assifier, 1	Bagging.	random fo	orests, Boos	sting: AdaBoost.	
Ì	, i i i i i i i i i i i i i i i i i i i	Gradient Boo	osting. XC	GBoost	,			,	0	

UNIT-IV Dimensionality Reduction: PCA, Kernel PCA (Textbook 4), LDA (08 Hrs) Model Evaluation and Optimization: Cross Validation, Grid Search, Regularization VINIT-V Multilayer perceptron's: Going forwards, Going backwards: Backpropagation of error, Multilayer perceptron in practice, Examples of using MLP. Reinforcement Learning: Overview, Example, Markov Decision Process, Uses of Reinforcement Learning: Overview, Example, Markov Decision Process, Uses of Reinforcement Learning: TEXTBOOKS: 1 1. Introduction to Machine Learning, Alpaydin E, MIT Press (2014) 3rdEdition 2 Machine Learning: The art and science of algorithms that make sense of data, Peter Flach, Cambridge, 2012 3 Machine Learning: An algorithmic perspective, Stephen Marsland, 2nd edition, CRC press, 2014. 4 Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow, Aurélien Géron, 2nd Edition, O'Reilly Publications, 2019 REFERENCE BOOKS: 1 1 The elements of statistical learning, Data Mining, Inference and Prediction, Trevor Hastie, Robert Tibshirani, Jerome Friedman, Second edition, 2012, Cengage. 3 Python Machine Learning: Machine Learning and Deep Learning with Python, scikit-learn, Tensorflow, Sebastian Raschka, Vahid Mirjalili, Second edition, 2020 Online MOC Courses: 1 1 "Machine Learning Course by Andrew Ng on Coursera 2 <t< th=""><th></th><th></th><th></th></t<>										
(08 Hrs) Model Evaluation and Optimization: Cross Validation, Grid Search, Regularization VINIT-V Neurons, NNs, Linear Discriminants: The Neuron, Neural Networks, The perceptron, Multilayer perceptron's: Going forwards, Going backwards: Backpropagation of error, Multilayer perceptron in practice, Examples of using MLP. Reinforcement Learning: Overview, Example, Markov Decision Process, Uses of Reinforcement Learning: Overview, Example, Markov Decision Process, Uses of Reinforcement Learning: Multilayer perceptron: The art and science of algorithms that make sense of data, Peter Flach, Cambridge, 2012 3 Machine Learning: The art and science of algorithms that make sense of data, Peter Flach, Cambridge, 2012 4 Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow, Aurelien Géron, 2nd Edition, O'Reilly Publications, 2019 REFERENCE BOOKS: 1 1. The elements of statistical learning, Data Mining, Inference and Prediction, Trevor Hastie, Robert Tibshirani, Jerome Friedman, Second edition, Springer, 2009. 2. Machine Learning: Machine Learning and Deep Learning with Python, scikit-learn, Tensorflow, Sebastian Raschka, Vahid Mirjalili, Second edition, 2020 00line MOOC Courses: 1 1. "Introduction to Machine Learning (IITKGP)" by Prof. Sudeshna Sarkar, on Swayam 3. "Machine Learning: Coxkbook-Practical Solutions from Preprocessing to Deep Learning, Chris Albon, Oreilly, 2018. 1. "Linear Discriminant Analysis", https://sebastianraschka.com/Articles/2014	UNI	T-IV	Dimensionality Reduction: PCA, Kernel PCA (Textbook 4), LDA							
Neurons, NNs, Linear Discriminants: The Neuron, Neural Networks, The perceptron, Multilayer perceptron's: Going forwards, Going backwards: Backpropagation of error, Multilayer perceptron in practice, Examples of using MLP. Reinforcement Learning: Overview, Example, Markov Decision Process, Uses of Reinforcement Learning: Overview, Example, Markov Decision Process, Uses of Reinforcement Learning: TEXTBOOKS: 1 1 Introduction to Machine Learning, Alpaydin E, MIT Press (2014) 3rdEdition 2 Machine Learning: The art and science of algorithms that make sense of data, Peter Flach, Cambridge, 2012 3 Machine Learning: An algorithmic perspective, Stephen Marsland, 2nd edition, CRC press, 2014. 4 Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow, Aurélien Géron, 2nd Edition, O'Reilly Publications, 2019 REFERENCE BOOKS: 1 1. The elements of statistical learning, Data Mining, Inference and Prediction, Trevor Hastie, Robert Tibshirani, Jerome Friedman, Second edition, 2012, Cengage. 2. Machine Learning: Machine Learning and Deep Learning with Python, scikit-learn, Tensorflow, Sebastian Raschka, Vahid Mirjalili, Second edition, 2020 0nline MOOC Courses: 1 1. "Machine Learning Course by Andrew Ng on Coursera 2. "Mathine Learning Course by Andrew Ng on Coursera 3. "Machine Learning Cookbook-Practical Solutions from Preprocessing to Deep Learning, Chris Albon, Oreill	(08	Hrs)	Model Evaluation and Optimization: Cross Validation, Grid Search, Regularization							
Neurons, NNs, Linear Discriminants: The Neuron, Neural Networks, The perceptron, Multilayer perceptron's: Going forwards, Going backwards: Backpropagation of error, (10 Hrs) Multilayer perceptron's: Going forwards, Going backwards: Backpropagation of error, Multilayer perceptron in practice, Examples of using MLP. Reinforcement Learning: Overview, Example, Markov Decision Process, Uses of Reinforcement Learning: TEXTBOOKS: 1 Introduction to Machine Learning, Alpaydin E, MIT Press (2014) 3rdEdition 2 Machine Learning: The art and science of algorithms that make sense of data, Peter Flach, Cambridge, 2012 3 Machine Learning: An algorithmic perspective, Stephen Marsland, 2nd edition, CRC press, 2014. 4 Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow, Aurélien Géron, 2nd Edition, O'Reilly Publications, 2019 REFERENCE BOOKS:										
UNIT-V (10 Hrs) Multilayer perceptron's: Going forwards, Going backwards: Backpropagation of error, Multilayer perceptron in practice, Examples of using MLP. Reinforcement Learning: Overview, Example, Markov Decision Process, Uses of Reinforcement Learning TEXTBOOKS: . 1. Introduction to Machine Learning, Alpaydin E, MIT Press (2014) 3rdEdition 2 Machine Learning: The art and science of algorithms that make sense of data, Peter Flach, Cambridge, 2012 3 Machine Learning: An algorithmic perspective, Stephen Marsland, 2nd edition, CRC press, 2014. 4 Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow, Aurélien Géron, 2nd Edition, O'Reilly Publications, 2019 REFERENCE BOOKS: . 1. The elements of statistical learning, Data Mining, Inference and Prediction, Trevor Hastie, Robert Tibshirani, Jerome Friedman, Second edition, Springer, 2009. 2. Machine Learning: Machine Learning and Deep Learning with Python, scikit-learn, Tensorflow, Sebastian Raschka, Vahid Mirjalili, Second edition, 2020 Online MOOC Courses: . 1. "Introduction to Machine Learning (ITKGP)" by Prof. Sudeshna Sarkar, on Swayam 3. "Machine Learning Cookbook-Practical Solutions from Preprocessing to Deep Learning, Chris Albon, Oreilly, 2018. 2. "Introduction to Machine Learning Cookbook-Practical Solutions from Preprocessing to Deep Learning, Chris Albon, Oreilly, 2018. 3. <			Neurons, NNs, Linear Discriminants: The Neuron, Neural Networks, The perceptron,							
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 4. foldcross-validation/ 5. Grid search for model tuning", https://medium.com/analyticsvidhya/illustrative-example- ofprincipal-component-analysis-pca-vs-lineardiscriminant-analysis-lda-is-105c431e8907 6. "Regularization in Machine Learning", https://towardsdatascience.com/regularization-inmachine- learning76441ddcf99a 		"A ge	entle introduction to K-fold cross-validation", https://machinelearningmastery.com/k-							
 5. Grid search for model tuning", https://medium.com/analyticsvidhya/illustrative-example- ofprincipal-component-analysis-pca-vs-lineardiscriminant-analysis-lda-is-105c431e8907 6. "Regularization in Machine Learning", <u>https://towardsdatascience.com/regularization-inmachine-</u> learning76441ddcf99a 	4.	foldere	oss-validation/							
 ofprincipal-component-analysis-pca-vs-lineardiscriminant-analysis-lda-is-105c431e8907 "Regularization in Machine Learning", <u>https://towardsdatascience.com/regularization-inmachine-learning76441ddcf99a</u> 	~	Grid	search for model tuning", https://medium.com/analyticsvidhya/illustrative-example-							
6. "Regularization in Machine Learning", <u>https://towardsdatascience.com/regularization-inmachine-</u> learning76441ddcf99a	5.	ofprine	cipal-component-analysis-pca-vs-lineardiscriminant-analysis-lda-is-105c431e8907							
0. learning76441ddcf99a	-	"Regu	larization in Machine Learning", https://towardsdatascience.com/regularization-inmachine-							
Iourning / 0 / / I duol / / u	6.	<u>lear</u> nir	1g76441ddcf99a							

	Code	Category	L	Т	Р	С	I.M	E.M	Exam	
B20	CS3104	PE	3			3	30	70	3Hrs.	
		•								
			ARTI	FICIAI	L INTEL	LIGEN	CE			
			(C	ommon	to CSE	& CSD)				
Cour	Course Objectives: Students are expected to									
1.	Know the	methodolog	y of Proble	em solvi	ng					
2.	Implemen	t basic AI alg	gorithms							
3.	Design ar	d carry out a	n empirica	l evolut	ion of dif	fferent alg	gorithms	on a problem	formalization	
Cour	se Outcon	nes: Students	will be ab	le to						
S No				Out	come				Knowledge	
5.110				04					Level	
1.	Summariz	ze different A	I problem	s, charac	cteristics	and state	space rep	presentation	K2	
2.	Apply he problems	uristic, unin	formed an	nd infor	med sea	irch strate	egies for	solving Al	K3	
3.	Apply Al uncertain	problem sol	ving appro	baches t	o propos	itional an	d predica	ate logic and	К3	
4.	Interpret	he given fact	s to differe	ent knov	vledge re	presentati	ional sche	emes, Prolog	K3	
5.	Summariz	e NLP, Plan	ning and d	evelopn	nent of E	xpert syst	ems.		K3	
	1	E 7/		1210						
		North Contraction	- 1	SY	LLABU	S	LULL	EGE		
	Int	oduction to	Artificia	l Intelli	gence: v	what is A	rtificial	Intelligence,	AI Problems	
UNI	Γ-I Def	ining the Pro	oblem as	a State	Space So	earch-Wa	ter jug p	roblem, 8-p	ızzleproblem,	
(10H	(rs) Tra	Travening Salesmen Problem, Turing Test, Production Systems, Problem								
	Cha	racteristics, i	roduction	System	Characte	ensues.				
	Sea	rch Technia		istic Se	arch Tec	hniques	Generate	-And-Test F	Iill Climbing	
UNI	Γ-II Pro	Problem Reduction AO*Algorithm Constraint Satisfaction Means Ends Analysis Un								
(10H	(rs) info	ormed searc	h-BFS &	DFS,	Informe	d Search	algorit	hms- Best-F	FirstSearch &	
× ·		Algorithm.		,			8			
	I									
	Syn	nbolic Logic	: Propositi	onal Lo	gic, First	Order Pr	edicate L	ogic, WFF r	epresentation,	
UNIT	con	version Alg	orithms	Clause,	Unifica	tion algo	orithm,	Resolution	algorithm in	
(10H	[rs] proj	positional log	ic and Res	solution	algorithr	n in predi	cate logio	2.		
(101	Rea	Reasoning under Uncertainty: Introduction to Non-Monotonic Reasoning, Truth								
	Ma	ntenance Sys	stems.							
						D	1 1 7 7		77 1 1	
	-1V Kn	owledge Rej	presentati	on usir	ng Rule	s: Procee	iural Vs	Declarative	Knowledge,	
(10H	rs) For	ward Vs Bacl	kward Rea	soning,	Matching	g, kete l	vlatching	•		

		PROLOG introduction, syntax and basic fields, key features, running queries, creating							
	and updating prolog database, lists in prolog, Cut & Fail in prolog, advantage								
	disadvantages, applications of prolog. Bayesian Networks, Fuzzy Logic, Fuzzy Se								
	Crisp Sets, Fuzzy inference and Fuzzy System. Dempster-Shafer theory.								
		Structured Representations of Knowledge: Semantic Nets, Partitioned Semantic Nets,							
	Frames, Conceptual Dependency and Scripts.								
		Planning: Components of a Planning System, Goal Stack Planning, Nonlinear planning,							
		hierarchical planning.							
		Natural Language Processing: Steps in the Natural Language Processing, Augmented							
UNI	Г-V	Transition Nets.							
(10 H	Irs)	Experts Systems: Overview of an Expert System (Examples, characteristics,							
		advantages and limitations, applications, why expert system) Architecture of Expert							
		Systems, Different Types of Expert Systems, Rule based expert system, Frame based,							
		and Decision Tree based expert system.							
Text]	Book	S:							
1	Artif	icial Intelligence, Elaine Rich and Kevin Knight, TataMcGraw-Hill Publications,							
1.	3 rd Ec	lition, Year-2010							
ſ	Intro	duction To Artificial Intelligence & Expert Systems, Patterson, PHI publications, First							
Ζ.	Editi	on, Year-2015							
3.	Artif	icial Intelligence- Saroj Kaushik, CENGAGE Learning							
Refere	ence E	Books:							
1.	Artif	icial Intelligence, George FLuger, Pearson Education Publications, 5th Edition, Year-2008							
r	Artif	icial Intelligence: A modern Approach, Russell and Norvig, Printice Hall, 3 rd Edition,							
Δ.	Year	-2015							
3.	Artif	icial Intelligence, Robert Schalkoff, McGraw-Hill Publications, 3rdEdition, Year-2002							
4	Artif	icial Intelligence and Machine Learning, Vinod ChandraS.S., Anand Hareendran S, First							
4.	Editi	on, Year-2014							

	Code		Category	L	Т	Р	С	I.M	E.M	Exam
B20	CD31	102	PE	3	0	0	3	30	70	3 Hrs.
					1			•	· ·	
			OBJ	ECT ORI	ENTED	ANALY	SIS ANI) DESIG	N	
					(F	or CSD)				
Cours	Course Objectives:									
1.	Und	lerstan	d the fundam	nentals of c	bject mo	odeling				
2.	. Understand and differentiate Unified Process from other approaches									
3.	Design with static UML diagrams									
4.	Design with the UML dynamic and implementation diagrams									
5.	Imp	rove tl	he software d	lesign with	design j	patterns				
Cours	se Ou	tcome	es: At the end	d of the cou	urse, Stu	dents wil	l be able t	0		
S No				6						Knowledge
5. INO				C	ourse O	outcomes				Level
1.	Illus	strate s	oftware deve	elopment p	rocess					K3
2.	Illus Diag	strate grams	Object Orie	ented Stru	ctural N	Iodeling	and Des	sign Clas	ss & Object	К3
3.	Des	ign Ba	sic Behavior	al Modelir	ngs.					K4
4.	Des	ign Ac	lvanced Beha	avioral Mo	deling					K4
5.	Des	ign a s	ystem by sel	ecting suit	able patt	ern.				K4
				E	IGIN	IEER	ING (OLL	EGE	
			Estd. 1980		SY	LLABUS	NOMO	US		
UNI	Г-І	Intro	duction to l	U ML : Imp	ortance	of mode	ling, prin	ciples of	modeling, ob	ject oriented
(12 H	(rs)	mode	ling, concept	ual model	of the U	ML, Arcl	nitecture,	Software	Development	Life Cycle.
		Basic	Structural I	Modeling:	Classes,	, Relatior	iships, cor	nmon Me	echanisms, and	l diagrams.
UNIT	'-II	Adva	nced Struct	ural Mod	eling: A	Advanced	classes,	advance	d relationship	s, Interfaces,
(10 H	(rs)	Types Close	δ and Koles, i δ Object	Packages.	• Torms	concen	te model	ing tech	vigues for Cla	es & Object
		Diagr	ams	Diagrams	• 1011115	, concep	is, model	ing teem	ilques for ele	iss & Object
		Diagi								
UNIT	'-III	Basic	Behavioral	Modeling	Interact	tions. Inte	eraction di	iagrams.		
(10 H	Irs)	Basic	Behavioral	Modeling	II: Use	cases, Us	e case Dia	agrams, A	Activity Diagra	ums.
		Adva	nced Behav	vioral Mo	deling:	Events	and signa	ls, state	machines, p	rocesses and
UNIT	-IV	Threa	ds, time and	space, state	e chart d	iagrams.				
(8 Hı	rs)	Archi	itectural Mo	deling: Co	omponer	nt, Deploy	yment, Co	omponent	diagrams and	Deployment
		diagra	ums.							

UNIT- (10 Hr	Case Study: The Unified Library application, College Management.							
ТЕХТВ	OOKS:							
1	Grady Booch, James Rumbaugh, Ivar Jacobson: The Unified Modeling Language User Guide,							
1.	Pearson Education.							
2	Hans-Erik Eriksson, Magnus Penker, Brian Lyons, David Fado: UML 2 Toolkit, WILEY							
۷.	reamtech India Pvt. Ltd.							
Referen	ce Books:							
1.	Meilir Page-Jones: Fundamentals of Object Oriented Design in UML, Pearson Education.							
2.	Pascal Roques: Modeling Software Systems Using UML2, WILEY- Dreamtech India Pvt. Ltd							
3.	Atul Kahate: Object Oriented Analysis & Design, The McGraw-Hill Companies.							
4	Object-Oriented Analysis and Design with the Unified Process By John W. Satzinger, Robert							
4.	B Jackson and Stephen D Burd, Cengage Learning.							
e-Reso	urces:							
1.	https://www.tutorialspoint.com/design_pattern/design_pattern_quick_guide.html							



(CodeCategoryLTPCI.ME.MExam									
B20	B20CD3103 PE 3 0 0 3 30 70								3 Hrs.	
HUMAN COMPUTER INTERACTION										
	(For CSD)									
Course Objectives:										
1.	1. Outline the importance of human computer interaction for a good design									
2.	Develop a	GUI application	ion for Ur	derstan	ding of U	sers.				
3	Distinguis	h Online Vs	Paper do	cumenta	ation in	various d	evelopme	ent processe	s and social	
5.	networkin	g.								
4.	Analyze so	creen design o	of various	applicat	ions in G	UI and W	'eb			
5.	Compare l	Device based	and Scree	n based	controls					
6.	Summariz	e effective fee	dback gu	idance a	nd assista	ance				
Cours	e Outcome	s: At the end	of the cou	rse Stud	ents will	be able to)			
S. No			Co	urse Ou	itcomes				Knowledge	
		ALL AND							Level	
1.	Understar	id the importa	nce of use	er interfa	ice and g	raphics-ba	ised syste	ems.	K2	
2.	Plan scree needs.	en designs wh	nich are f	ast and	pleasing	composit	ions to s	atisfy user	K3	
3.	Understar	nd business fur	nctions an	d desigr	n standaro	ls or style	guides	GE	K2	
4.	Understar screens.	d various so std. 1980	creen cor	nponent	s and d	etermine	the nav	igation of	K2	
5.	Summariz	e various scre	en device	s and sc	reen-base	ed control	compone	ents.	K2	
				SYL	LABUS					
	The	User Interface	e: Introdu	uction,	Importan	ce of the	User Ir	nterface, Im	portance and	
UNIT	'-I benef	its of Good	Design, H	History of	of Huma	n Compu	ter Interf	face The G	raphical User	
(10H	rs) Interf	ace: popularit	y of grap	hics, the	e concep	t of direc	t manipu	lation, grap	hical system,	
	Characteristics, Web user – Interface popularity, characteristics- Principle							ples of user		
	Interi	ace.								
	The 1	Iser Interface	Design	Process	Obstacl	es and Pi	itfall in t	he developr	nent Process	
UNIT-II Usability The Design Team Human Interaction with Computers Principles							ples of User			
(10.Hrs) Interface Design Internation Human Characteristics in Design. Human Consideration							nsideration in			
	Desig	;n	1				C .			
	1									
UNIT	Unde Unde	rstanding Bu	siness F	unctions	: Busine	ess Defin	itions &	Requirem	ent analysis,	
(10H)	rs)	mining Busin	ess Funct	ions, De	sign stan	dards or S	Style Gui	des, System	Training and	
	Documentation									

UNI	Principles of Good Screen Design: Human considerations in screen De design. goals, test for a good design, Technological considerations in IntIT-IVSystem Menus and Navigation Schemes: Structure, Functions, Contex	sign, interface erface Design. tt, Formatting,					
(10 I	(Hrs) Phrasing .and Selecting, Navigating of Menus, Kinds of Graphical Me	nus. Windows					
	Interface: Windows characteristic, Components of Window, Windows	s Presentation					
	Styles, Types of Windows, Window Management, Web systems.						
UNI	IT-V Device and Screen-Based Control: Device based controls, Operable (entry/read-only Controls, Section Controls, Combining Entry/Selection C	Controls, Text Controls, Other					
(12 ł	(12 Hrs) Operable Controls and Presentation Controls, Selecting proper controls.						
Text	t Books:						
1.	The Essential Guide to User Interface Design, Wilbert O. Galitz, Wiley India Edi	ition.					
2.	Sharps Interaction Design, Prece, Rogers, Wiley India						
Refer	erence Books:						
1.	Designing the user interfaces, Ben Shneidermann, 3rd Edition, Pearson Education	n Asia					
2.	User Interface Design, Soren Lauesen, Pearson Education						
3.	Essentials of Interaction Design, Alan Cooper, Robert Riemann, David Cronin						
4	Human Computer Interaction, Alan Dix, Janet Fincay, GreGoryd, Abowd, F	Russell, Bealg					
4.	+. Pearson Education						
e-Res	esources:						
1.	https://onlinecourses.nptel.ac.in/noc18_cs23 ERING COLLEGE						
2.	https://www.scribd.com/interest/Human-Computer-Interaction/docs						
3.	https://www.interaction-design.org/courses/human-computer-interaction						

(CodeCategoryLTPCI.ME.MExam										
B20CD3104 PE 3 0 0 3 30 70									3 Hrs.		
VISUAL DESIGN AND COMMUNICATION											
	(For CSD)										
Course Objectives:											
1.	1. Students will develop the ability to create visual compositions using basic elements and apply appropriate principles of visual composition to communicate ideas.										
2.	2. Students will begin to understand the visual language and develop the ability to perceive, visualize and communicate using visual narratives.						to perceive,				
3.	Students Typograph	will develop ny, Photograp	the abil hy and Vic	ity to a leograph	apply the y.	e visual o	dynamics	s of visual	language in		
4.	4. Visualization process to evolve mental imageries that represent solutions to simple communication problems										
5.	Students w	vill be able to	execute de	sign sol	utions usi	ng approp	riate soft	ware program	imes		
Cours	e Outcome	s: At the end	of the cour	se Stude	nts will b	e able to					
S. No	1			Outco	me				Knowledge Level		
1.	Understan	d the visual la	anguag <mark>e an</mark>	d its eler	nents				K2		
2.	Understan	d the principl	es of visua	l langua	ge	NG C	OLLE	GE	K2		
3.	Understan communic	d the fundam ation	entals of ty	pograph	y and its	application	n as a me	ans of visual	K2		
4.	Use Photo use it for v	ography and Visual commu	Videograpl nication	ny to cre	eate conte	ent (galler	y & shor	rt video) and	К3		
5.	Understan solving an	ding commur d communica	nication the ation	eories, st	orytelling	and other	^r concepts	s for problem	K2		
				SYI	LABUS						
	INTRO	DUCTION	TO VIS	UAL D	ESIGN:	The imp	ortance	of understan	nding visual		
UNIT	I languag	ge and its rela	tion in con	text to n	ature and	environm	ent.				
(10 Hr	s) ELEM	ENTS OF V	ISUAL LA	ANGUA	GE: Exp	loring and	understa	nding Dots, L	Lines, Forms,		
	Space,	Pattern, Textu	are and Co	lour as a	n element	of visual	language	•			
						~ ~ ~					
		DUCTION	то тн	E PRI	NCIPLE	S OF	VISUAL	LANGUA	GE: Visual		
UNIT-	II explora	tions and ex	periments	with Fo	orm, Colo	our, and S	Space, To	exture, in rel	ation to the		
(10 Hr	(s) context	and environ	nments –	Concept	s of har	mony, bal	lance, co	ntrast, propo	ortion, order,		
	symme	try, asymmet	try, rhythn	n, tensio	on, juxtaj	position, p	proximity	, size, scale,	proportion,		
	oriental	non, angnine	int, variety,	grauatio	лі, uomm		Junation	i, italisitioli el			

L

	INTRODUCTION TO FUNDAMENTALS OF TYPOGRAPHY: Introduces Typography									
	as a means of Communication and engages in typographical explorations to understand the									
UNIT	-III technicalities, nuances and aesthetics of types. Study of visual principles of text and image									
(10 H	0 Hrs) composition: Layouts, Grids, Content Development and Information Hierarchy. Applica									
of Typography, Image and layouts in the design of signage systems, identity syste										
	communications,									
	INTRODUCTION TO PHOTOGRAPHY: Study of photography as a medium to									
	document, communicate and create photographic imagery. Exploring photo story as a									
UNIT	-IV narrative medium									
(10 H	INTRODUCTION TO VIDEOGRAPHY . Study of videography as a medium to document									
	communicate and create a short 2 minute video									
	communicate and create a short 2 minute video.									
	COMMUNICATION THEODIES SEMIOTICS AND VISUAL DEDCEDTION.									
	COMMUNICATION THEORIES, SEMIOTICS AND VISUAL PERCEPTION:									
TINIT	Understand the process of communication and the theories that make a difference to the									
	-v development of a visual language.									
(10 H	Irs) STORY TELLING, NARRATIVES AND ITS ROLE IN DESIGN: Understand									
	storytelling and narratives as effective methods to scope problems and problem solving									
	processes.									
Text	Books:									
1	Walls chlaeger, Charles, & Busic - Synder, Cynthia, Basic Visual Concepts and Principles for									
1.	Artists, Architects and Designers, McGraw-Hill, 1992.									
Refe	rences:									
1	Buxton, Bill, Sketching User Experience: Getting the Design Right and the Right Design									
1.	(Interactive Technologies), Morgan Kaufmann, 2007.									
2	Caplin, Steve; Banks, Adam, The Complete Guide to Digital Illustration, Publisher: Watson -									
2.	Guptill Publications, 2003									
3	Demers, Owen, Digital Texturing & Painting, Publisher: New Riders Press; Bk & CD-ROM									
5.	Edition,2001.									
4.	Cairo, Alberto, The Functional Art. New Riders, 2013.									

C	CodeCategoryLTPCI.ME.MExam										
B20A	B20AM3104 PE 3 0 0 3 30 70										
	INTERNET OF THINGS										
	(Common to AIML & CSD)										
Pre-re	Pre-requisites: Computer Networks										
Course	Course Objectives: Students are expected										
1.	1. To understand building blocks of IoT and their characteristics										
2.	To Know	w various archi	tectures	and proto	cols in I	oT and se	curity issue	8			
3.	To use c	loud services f	or data a	nalytics in	n IoT ap	plication	8				
4.	To deve	lop IoT applica	tions usi	ng Ardui	no progr	amming.					
Course	Outcom	es: At the end	of the co	urse stude	ents will	be able t	0				
S No				Oute	ome				Knowledge		
5.110				Out	ome				Level		
1.	Discuss	various Design	s of IoT	and IoT a	architect	ures			K2		
2.	Illustrate	e various comm	nunicatio	n protoco	ls in IoT				K3		
3.	Use of	various sensor	s and A	ctuators i	in IoT a	applicatio	ns and Imp	lement IoT	K3		
	applicati	ons using Ard	lino.		_						
4.	Analyse	data in IoT ap	plication	S.					K4		
5.	Analyse	various securit	ty issues	IoT appli	cations.	ING	COLLE	GE	K4		
		Fetd 1980			AUTO	NOMO	JUS				
		Lotarityoo		SYI		S					
	Int	roduction to 1	nternet	of Thing	s: Defin	Ition &C	haracteristic	s of lot, Phy F Eurotional	Ploake IoT		
UNIT	$[-\mathbf{I}] \begin{bmatrix} 0 & \mathbf{I} \\ \mathbf{C} \end{bmatrix}$	munication M	101, 101	ADI's Io'	5, LOGIC T lovala	and depl	1 01 101 - 101	l runcuonal	DIOCKS, 101		
(10 H	rs) $\begin{bmatrix} con \\ IoT \end{bmatrix}$	IoT Network Architecture and Design: Drivers Behind New Network Architectures.									
	Cor	nparing IoT A	rchitectu	res. A Sin	nplified	IoT Arch	itecture.		licinteetures,		
		1 0		,	r						
	Co	mmunication	Techn	ologies:	wired	Comm	unication	Technologie	s, wireless		
UNIT	-II Con	nmunication T	echnolog	gies.				C			
(10 H	rs) IoT	Access Techr	nologies:	PHY/MA	AC Laye	r (IEEE 8	802.15.4), Lo	DRAWAN, F	RPL.		
	Message Communication Protocols for Connected Devices - CoAP, XMPP, MQTT.								PP, MQTT.		
	ΙΟ	Г Physical dev	rices and	Endpoir	nts: Basi	c buildin	g blocks of a	an IOT devic	e.		
	Ser	sors, Particip	patory s	sensing,	RFIDs:	Sensor	Technology	y, Participato	ory sensing,		
UNIT	-III Ind	ustrial IOT	and Au	tomotive	IOT,	Actuator	, Radio F	Frequency I	dentification		
(10 H	rs) tech	nology.	.41. 4 7	•	- 4	C A 1 '	C		1		
	Pro	gramming w	ith Ard	uino: Fe	atures o	or Arduir	Troffic corr	trol outtom	uino board,		
	Arc	iumo ide, pro	grammin	ig Elemer	ns, Case	sindles:	rranne con	uoi system,	DULT Sensor		

		with Arduino.								
UNIT (8 H	 (8 Hrs) Data Acquiring, Organising, Processing and Analytics: Introduction, Data Acquiring and storage, Organising the Data, Transaction, Business Processes, Integration and Enterprise Systems, Analytics, Knowledge Acquiring, Managing and Storing Processes. 									
UNIT-V (8 Hrs) IoT Privacy, Security and Vulnerabilities Solutions: Vulnerabilities, Security (8 Hrs) IoT Privacy, Security and Vulnerabilities Solutions: Vulnerabilities, Security Requirements and Threat Analysis, IoT Security Tomography and Layered Attack model, Identity management and establishment, Access control secure messa communication, Security models, profiles and protocols for IoT.										
		Case studies mustrating for Design. Tione Automation, Environment, Agriculture								
TEX	ГВОС)K:								
1.	1. Internet of Things - A Hands-on Approach, Arshdeep Bahga and Vijay Madisetti, Universities Press, 2015.									
2.	IoT Thin 2475	Fundamentals: Networking Technologies, Protocols, and Use Cases for the Internet of gs - David Hanes, Gonzalo Salgueiro, Patrick Grossetete Robert Barton, Jerome Henry. O Copyright© 2017 Cisco Systems, Inc. Published by: Cisco Press 800 East 96th Street.								
3.	Inter priva	net of Things: Architecture and Design Principles by Raj Kamal, McGraw Hill Education te limited, 2017.								
4.	Inter	net of Things, Jeeva Jose, Khanna Publishing; First edition (2018).								
REFI	EREN	CE BOOKS:								
1.	Desi	gning the Internet of Things, Adrian McEwen and Hakim Cassimally, Wiley, 2013								
2.	Getti	ng Started with the Internet of Things Cuno Pfister, Oreilly, 2011								
3.	3. Getting Started with Raspberry Pi, Matt Richardson & amp; Shawn Wallace, Reilly (SPD), 2014.									
E-R	esour	ces:								
1.	Intro	duction to Internet of Things, https://swayam.gov.in/nd1_noc20_cs66/preview								
2.	An I <u>http</u>	ntroduction to Programming the Internet of Things(IoT) specialization, <u>os://www.coursera.org/specializations/iot</u>								

C	Code Category L T P C I.M E.M Exam										
B200	B20CS3108 PC 0 0 3 1.5 15 35		35	3 Hrs.							
				DAT	TA MININ	NG LAB					
				(Comm	non to CS	E & CSD)				
Cour	Course Objectives: Students are expected to learn										
1	The basic statistical measures of data mining in order to prepare for real-world problems										
2	The ability to apply various data mining algorithm										
3	Package	es and librari	es of R a	and also f	familiar wi	th functio	ns used i	in R for visua	alization		
4	Using F	environmer	it to con	duct anal	ytics on la	rge real lif	fe datase	ts			
Cours	se Outco	mes: At the	end of th	e course	students v	vill be able	e to				
S. No				Oı	itcome				Knowledge		
1	D 1	.1	1		11	1			Level		
1	Extend	the functiona	ulity of k	t by using	g add-on p	ackages		vanious data	K3		
2	manipu	le data from	n them	and oth	er source	es and pe	eriorin	various data	K4		
2	Use R (Graphics and	Tables	to apply	and visual	ize results	of vario	ous statistical	- V2		
3	operatio	ons on data							K.J		
4	Apply t	he <mark>kn</mark> owledg	e of R g	ained to a	data analyt	tics and da	ta minin	g of real-life	К3		
	datasets		87								
	j		97—	ENG	SYLLAB	US	COL	FGE			
1	Impler	nent all basic	R comr	nands.		TONOM					
2	Interac	t data throug	h .csv fi	les (Impo	ort from an	d export t	o .csv fil	es).			
2	Get an	d Clean data	using s	wirl exer	cises. (Use	e 'swirl' p	ackage,	library and i	nstall that topic		
3	from s	wirl).	e		×	Ĩ	U ,		1		
4	Visual	ize all Statis	tical me	asures (N	Aean, Moo	de, Media	n, Range	e, Inter Quar	tile Range etc.,		
	Create	a data frame	with the	s and Sca s followir	ng structur						
	a.	EMP ID	EMP	NAME	SALARY	STA	RT	DATE			
	b.	1	Satish		5000	01-11	-2013				
	c.	2	Vani		7500	05-06	5-2011				
_	d.	3	Rame	sh	10000	21-09	9-1999				
5	e.	4	Pravee	en	9500	13-09	9-2005				
	f.	5	Pallav	i	4500	23-10)-2000				
	a.	Extract two c	olumn r	names usi	ing columi	n name.					
	b.	Extract the fi	rst two 1	rows and	then all co	olumns.					
	c.	Extract 3rd a	nd 5th r	ow with 2	2nd and 4t	h column.					
	Write	R Program us	sing 'app	oly' grou	p of functi	ons to cre	ate and a	apply normal	ization function		
6	on eacl	h of the nume	eric varia	ables/colu	umns of in	is dataset f	to transfo	orm them int	0		
	a. h	u to 1 range		i-max noi		II.					
	b. a value around 0 with z-score normalization.										

7	Create a data frame with 10 observations and 3 variables and add new rows and columns to it using 'rbind' and 'cbind' function
8	Write R program to implement linear and multiple regression on 'mtcars' dataset to estimate the value of 'mpg' variable, with best R2 and plot the original values in 'green' and predicted values in 'red'.
9	Implement k-means clustering using R.
10	Implement k-medoids clustering using R.
11	implement density based clustering on iris dataset.
12	implement decision trees using 'reading Skills' dataset.
13	Implement decision trees using 'iris' dataset using package party and 'rpart'.
14	Use a Corpus() function to create a data corpus then Build a term Matrix and Reveal word frequencies.
REF	ERENCE BOOKS:
1.	R and Data Mining: Examples and Case Studies, 1st ed, Yanchang Zhao, Sprnger, 2012.
2	R for Everyone, Advanced Analytics and Graphics, 2nd ed, Jared Lander, Pearson, 2018.
e-Res	sources:
1.	www.r-tutor.com



Co	CodeCategoryLTPCI.ME.MExam										
B200	CD3105	PC	0		3	1.5	15	35	3 Hrs.		
	ADVANCE DATA STRUCTURES LAB										
	(For CSD)										
Cours	Course Objectives:										
1.	1. To develop skills to implement linked list, trees, graphs.										
2	To gain experience with the implementation of basic Hash Tables and pattern matching										
2.	algorithm	18.									
3.	To acquir	re comprehensiv	e know	ledge of	advanc	ed data str	ructures an	d implemen	t the same in		
	different	software applica	tions.								
~											
Course	e Outcome	es: The students	will be a	able to							
S. No.			С	ourse Oi	itcome	S			Knowledge		
	Obtain li	inlead list and Te	lana lana	wladaa		ical applia	otiona main	a different	Level		
1.	language	inked list and 11	ees kno	wiedge i	n praci	ical applic	ations usin	g different	K3		
2	Impleme	s. ont graph algorith	me to e	olve vari	ous rea	l time annli	ications		K3		
2.	Impleme	ent different Hast	n Tables	and Patt	ern Ma	tching Alg	orithms		K3		
5.	mpieme		I I doite				onunns.		13		
LIST	OF EXPI	ERIMENTS									
1.	Impleme	entation of singly	linked	list	FFF	ริเพิร (TOLE	GE			
2.	Impleme	entation of doubl	v linked	list		<u>ANOMO</u>					
3.	Program	to reverse the no	odes in a	a circular	linked	list	Q Q				
4.	Program	to perform oper	ations o	n two po	lynomi	als using li	nked list				
5.	Impleme	ent traversal tech	niques i	n binary	tree	<u> </u>					
	Beginnir	ng with an empty	binary	search tr	ee, con	struct bina	ry search tr	ee by inserti	ng the values		
	in the or	der given. After	construc	cting a bi	nary tre	ee -	-	·	-		
	\checkmark	Insert new node									
6.	\checkmark	Find number of	nodes i	n longest	path						
	\checkmark	Minimum data v	value fo	und in th	e tree						
	\checkmark	Change a tree so	o that th	e roles of	the lef	t and right	pointers ar	e wrapped a	t every node.		
	\checkmark	Search a value									
	Write a p	program to perfo	rm the f	ollowing	operat	ions					
7.	\blacktriangleright	Insertion into an	a AVL-t	ree							
	\blacktriangleright	Deletion from a	n AVL-	tree							
	Program	to implement pr	iority q	ueue usir	ig Hear)					
8.	\blacktriangleright	Inserting new el	ement								
	\succ	Deletion of min	imum e	lement							
9.	Write a p	program to imple	ment D	FS and E	BFS trav	versals.					

10.	Write a program to find minimum spanning tree using Prim's Algorithm								
11.	Write a program to find the minimum spanning tree using Kruskal's Algorithm.								
12.	Write a program to implement topological sort.								
13	Write a program for creating an Open Addressing Hash Table with linear probing and quadratic								
13.	probing.								
14.	Write a program to implement Naive Pattern Matching algorithm								
15.	Write a program to identify the desired patterns with Knuth-Morris-Pratt (KMP) algorithm.								
16.	Write a program to implement the Rabin Karp pattern matching algorithm.								
Refere	nce Books:								
1	Data Structures using C by Aaron M. Tenenbaum, Y. Langsam and M.J. Augenstein, Pearson								
1.	Education, 2009.								
2.	Data Structures with C by Seymour lipschutz, Schaum Outline series, 2010.								
3.	Data Structures using C by R. KrishnaMoorthy G. Indirani Kumaravel, TMH, New Delhi, 2008.								
1	Data Structures and Algorithms: Concepts, Techniques and Applications – G.A.V.Pai, Tata Mc								
4.	Graw Hill Publishers								
5.	Advanced Data Structures – Peter Brass, Cambridge University Press, 2008								
Refere	nce links :								
1.	https://www.cs.usfca.edu/ ~ galles/visualization/Algorithms.html								
2.	http://cse01-iiith.vlabs.ac.in/List%20of%20experiments.html?domain=Computer%20 Science								



Co	CodeCategoryLTPCI.ME.MExam									
B20HS3102		SOC	1		2	2		50	3Hrs.	
	SOFT SKILLS									
	(Common to AIDS, AIML, CSBS, CSD, CSE, ECE, & IT)									
Course	Course Objectives:									
1.	To fam	iliarize students	s with so	ft skills a	and how	they influence	uence the	eir professio	onal growth.	
2.	To buil	d/refine the pro nfidence throug	fessional gh attitud	l qualitie e buildii	es/skills : ng.	necessary	for a pr	oductive car	reer and to in	
Cours	se Outcou	mes• At the end	l of the c	ourse sti	idents w	ill be abl	e to			
Cours		ines. At the end			udents w		c 10		Knowledge	
S.No				Outco	ome				Level	
1	Apply professi	soft skills in ional relationsh	the wo	ork plac ng infor	ce and med dec	build be isions.	etter per	rsonal and	К3	
2	 Participate in group discussions/group activities, exhibit team spirit, use language effectively according to the situation, respond to their interviewer/employer with a positive mind, make answers to the questions asked during their technical/personal interviews, exhibit skills required for the different kinds of interviews (stress, technical, HR) that they would face during the course of their recruitment process. 									
	- (H	A BA			1					
				SYL	LABUS					
1.	Introdu	ction to Soft Sk	cills, Sigr	nificance	e of Inter	& Intra-	Personal	Communic	ation	
2.	SWOT	Analysis, Crea	tivity & l	Problem	Solving	NOMO	lis			
3.	LSRW,	JAM, Presenta	tion Skil	ls						
4.	Buildin	g a positive att	itude, Lea	adership	& Tean	n Work				
5.	Goal Se	etting – Guideli	nes for C	Goal Sett	ing					
6.	Group I	Discussion: Ess	ential gu	idelines						
7.	Telepho	one Etiquette, T	elephoni	c Interv	iew					
8.	Resume	e Preparation: C	Common	resume	blunders	, tips for	betterme	ent, Resume	Review	
9.	Employ Social I	ability Skills: Entrepreneurshi	Emotion p, Stress	al Intelli Manage	gence, I ement.	Report W	riting, S	ocial Consc	iousness and	
10.	Awaren workpla	ess about Incace, Knowing a	lustry, C bout Sele	Companiection Pr	les, Imp rocess	ortance	of resea	arching the	prospective	
11.	Intervie	w Skills: Type	s of Inter	views, N	Mock Int	erview, I	Do's and	Don'ts of Ir	nterview.	
Text l	Books:									
1	Soft Sk Univers	cills & Employ sity Press India	yability Pvt. Ltd.	Skills b	y Samin	a Pillai	and Agi	na Fernande	ez, Cambridge	
2	Soft Sk	ills, by Dr. K. A	Alex, S. C	Chand &	Compa	ny Ltd., N	New Dell	hi		
Refer	ence Boo	ks:								
1	The Art	t of Public Spea	aking by	Dale Ca	rnegie					

2	The Leader in You by Dale Carnegie							
3	Emotional Intelligence by Daniel Golman							
4	Stay Hungry Stay Foolish by Rashmi Bansal							
5	I have a Dream by Rashmi Bansal.							
Additio	onal Materials							
1	https://www.youtube.com/watch?v=LTnI7cmpDZI							
2	https://www.youtube.com/watch?v=ic5O2sxhH9M							
3	https://www.youtube.com/watch?v=4ZQkYSpmOdU							
4	https://www.youtube.com/watch?v=d8p-5WcXoRs							
5	https://www.youtube.com/watch?v=yZOar04g4zk&t=94s							



(Code	Category	CodeCategoryLTPCI.ME.M					E.M	Exam	
B20 I	MC3105	MC			3				3 Hrs.	
			DIGI	FAL MA	RKETIN	G LAB				
(For CSD)										
Course Objectives:										
1	To famil	iarize students	to method	ologies, t	ools and te	echniques	involved in	n digital m	arketing.	
2	To provide students with specific knowledge and sufficient background that will allow them to									
2	pursue their careers in the digital marketing domain.									
3	To experiment and study how an online business is built and promoted digitally.									
4	To design and create ad campaigns on social media and other platforms.									
S. No	o Course Outcomes Knowledg									
	Level									
1	Ability to build a sample online business and promote using digital marketing K4									
2	Ability 1	to develop a	digital m	arketing	plan to	address c	ommon n	narketing	K4	
	A bility t	es use multi ch	annal mar	koting st	rotogias	cing SEO	Doid Ad	vortiging		
3	Social M	edia Mobile F	anner mai mail and o	other Dis	nlav Medi	a tools	, Falu Au	verusing,	K3	
4	Ability to	devise strateg	ies and stu	dv the in	pray mean	isiness obj	iectives.	<u> </u>	К3	
					1					
LIST	OF PRO	GRAMS	EN	IGIN	EERII	NG-CC)LLEG	E		
1	Create a	website and hos	st - Person	al / E-Co	mmerce /	Other				
2	Design a	n online busine	ss strategy	for pron	notion					
3	Impleme	nt Search Engir	ne Optimiz	zation (SI	EO) for yo	ur website				
4	Create an	Ad using Sear	ch Engine	Marketin	ng (SEM)					
5	Advertise	e using Social N	Aedia Mar	keting or	n Facebook	, Instagrai	m and You	itube		
6	Content I	Marketing - Cre	eate a sam	ple digita	l content i	n the form	of a blog	or multime	edia content	
7	Impleme	nt Email and W	/hatsapp /S	SMS mar	keting to p	otential cu	istomers			
8	Impleme	nting Web Ana	lytics on y	our webs	site and ob	serve the d	lata insigh	ts		
Refer	ences:									
1.	"Digital N	Iarketing All-I	n-One For	Dummie	es" by Step	hanie Dia	mond, Wil	ey Publica	tions, 2020	
2.	https://lea	rndigital.withg	oogle.com	digitalg	arage/cour	se/digital-1	marketing			



SAGI RAMA KRISHNAM RAJU ENGINEERING COLLEGE (AUTONOMOUS)

(Affiliated to JNTUK, Kakinada), (Recognized by AICTE, New Delhi) UG Programmes CE,CSE,ECE,EEE,IT & ME are Accredited by NBA, Accredited by NAAC with A⁺ CHINNA AMIRAM (P.O):: BHIMAVARAM :: W.G.Dt., A.P., INDIA :: PIN: 534 204

Regula	Regulation: R20				III / IV - B.Tech. II - Semester							
	COMPUTER	SCIENC	E AN	D DE	SIGN	1						
	SCHEME OF INSTRUCTION & EXAMINATION (With effect from 2021-22 admitted Batch onwards)											
Course Code	Course Name	Catego ry	Cr	L	Т	Р	Int. Marks	Ext. Marks	Total Marks			
B20AM3201	Computer Networks	PC	3	3	0	0	30	70	100			
B20CD3201	Automata theory and Compiler Design	PC	3	3	0	0	30	70	100			
B20AM3203	Software Engineering	PC	3	3	0	0	30	70	100			
#PE-II	Professional Elective -II	PE	3	3	0	0	30	70	100			
#OE-II	Open Elective-II	OE	3	3	0	0	30	70	100			
B20CD3204	Computer Networks Lab	PC	1.5	0	0	3	15	35	50			
B20AM3209	Algorithms for Efficient Coding Lab	PC	1.5	0	0	3	15	35	50			
B20CD3205	Internet Of Things Lab	PC	1.5	0	0	3	15	35	50			
B20CD3206	MERN Stack Technologies- Module I (Skill oriented course)	SOC	2	1	0	2		50	50			
B20MC3201	Employability Skills	MC	0	3	0	0						
B20HS3204	*Gender Sensitization	HS	0	2	0	0						
	1	TOTAL	21.5	21	0	11	195	505	700			

	Course Code	Course					
	B20CD3202	Deep Learning					
#PE-II	B20CD3203	NS (Cryptography & Network Security)					
	B20AM3206	Distributed Systems					
	B20CS3207	Network Programming					
#OF II	Student has to study one Open Elective offered by CE or ECE or EEE or ME or						
#OE-II	S&H from the list enclosed.						

*Note: Gender Sensitization is a Self-Learning noncredit Audit Course

(Code	Category	L	Т	Р	С	I.M	E.M	Exam	
B20	AM3201	PC	3			3	30	70	3 Hrs.	
			CON	APUTE	R NETW	ORKS				
(Common to AIML & CSD)										
Course Objectives: Students are expected										
1	To study t	the basic taxo	nomy and	terminol model	ogy of th	e compu	ter netwoi	rking and en	umerate the	
2	To study of	data link lave	concepts	design i	ssues an	d protoco	ols			
3	To study MAC laver Random Access Protocols I AN									
4	To gain ki	nowledge on	Network la	aver and	Routing	Algorith	ns.			
.5	To learn T	Fransport lave	r services.	and prot	ocols.					
6	To acquire	e knowledge (of Applica	tion lave	r protoco	ls				
	10 uoquii		or reperiou		r protoco					
Course	Outcomes	: At the end of	of the cour	se studer	ts will be	e able to				
e ourse				se staden					Knowledge	
S. No.	Outcome								Level	
1	Illustrate the OSI reference model, TCP/IP, and Digital transmission techniques									
2	Demonstrate Data Link Layer protocols K3									
3	Compare and contrast MAC protocols, various types of LANs K3									
4	Summariz	e various net	work layer	· services	and Rou	iting algo	rithms	:	K3	
5	Implemen	t Transport la	yer and ap	plication	layer pr	otocols		.uc	K3	
	E	std. 1980				a de la carde de	<i></i>			
				SYL	LABUS					
UNI7 (10 H	Introd reference (rs) MAN (Digi	duction: Intro ence model, 7 I, WAN). Ph tal-to-Digital,	oduction t TCP/IP ref sysical lay Analog-to	o Comp erence m er: Data p-Digital	uter Net odel. Ne and Sig), multipl	works, 1 etwork to gnals, Di lexing (F	Network pologies, gital sigr DM, TDN	Models (pro types of net nals, Digital A), Transmis	otocols): OSI works (LAN, transmission ssion media.	
UNIT (8 Hi	 UNIT-II (8 Hrs) Data Link Layer: Error Detection & Correction: types of errors, Error Detection (Parity CRC, Check Sum), Error Correction (Using hamming distance), Data Link Layer services framing, flow control, error control. Error & Flow control mechanisms: stop and wait, Go back N and selective repeat, High Level Data Link Control (HDLC). 								ection (Parity, ayer services: and wait, Go	
UNIT- (10 F	Medium access control: Random access: Aloha, Slotted Aloha, CSMA, CSMA/CD, and CSMA/CA, Local area networks: Ethernet, Types of ethernet (Token Ring, Fast Ethernet, Gigabit Ethernet), Personal Area Network: Bluetooth (Architecture), Wireless LANS: IEEE 802.11(Architecture, MAC sub layer).									

UNIT (8 Hi	NIT-IV 8 Hrs)Network layer: Network Layer Services, IPV4 Address, Subnetting, Super-nettin Classless addressing, Internet Protocol (IP, ARP, DHCP, ICMP), IPV6 Address formation Routing algorithms: Distance vector, Link state, Network Address Translation (NAT).									
	Transport layer: UDP (User Datagram, Services, Applications), TCP (TCP Services,									
UNIT	'-V features, Segment, Connection establishment and termination, sliding window, flow, and									
(8 Hı	rs) congestion control), Application Layer: Application Layer services and protocols									
	including www, DNS, SMTP, POP, FTP, Telnet, HTTP, Firewalls.									
TEXT	BOOK:									
1.	Data Communication and Networking, Behrouz A. Forouzan, McGraw Hill, 5th Edition, 2017.									
REFE	RENCE BOOKS:									
1.	Data and Computer Communications, William Stallings, Pearson, 10th Edition, 2013.									
2	Computer Networks, Andrew S. Tanenbaum, David J. Wetherall, Pearson Education India;									
۷.	5 th edition, 2013.									
2	Computer Networks: A Systems Approach, LL Peterson, BS Davie, Morgan-Kauffman, 5th									
5.	Edition, 2011.									
4	Computer Networking: A Top-Down Approach JF Kurose, KW Ross, Addison-Wesley, 5th									
4.	Edition, 2009.									



C	ode	Category	L	Т	Р	С	I.M	E.M	Exam		
B20	CD3201	PC	3	0	0	3	30	70	3 Hrs.		
AUTOMATA THEORY AND COMPILER DESIGN											
(For CSD)											
Course Objectives:											
1	1 To learn fundamentals of Regular and Context Free Grammars and Languages.										
2	To understand the relation between Contexts free Languages, PDA.										
3	To study the various phases in the design of a compiler.										
4	To unde	rstand the desig	gn of top-o	lown and	l bottom-	up parser	s.				
5	To understand syntax directed translation schemes and approaches to generate code for atarget machine.										
Cours	se Outco	nes: At the end	of the co	urse Stuc	lents will	be able t	0				
S No				Outcor	nes				Knowledge		
5.10	D .			Outcon					level		
1	Design an automata for given language and equality regular expressions K4										
2	Design	various parse tr	ees using	parsing a	lgorithm	s for the	given gran	nmar	K4		
3	Constru	ct various form	s of interr	nediate c	ode gener	ration.	6		К3		
4	Apply 1 storage	echniques to	generate	optimize	d and in	nprove p	erforman	Ce in code	K3		
5	Design	algorithms for c	ode gener	ration	AUTO	NOMO	US		K4		
				SYL	LABUS						
UNI' (10 H	T-I Irs)	pressions, Finit FA to DFA. Ap	ge and Re te Automa plications	gular Ex ata – DF of Finite	A, NFA.	S: Langu Convers ta to lexio	iages, Def ion of reg cal analysi	unition Lang gular expres	uages regular sion to NFA,		
	C	ntext Free gr	ammars	and nar	sing: Cor	ntext free	orammar	s derivation	n parse trees		
UNIT-IIContext Free grammars and parsing: Context free grammars, derivation, ambiguity LL(K) grammars and LL (1) parsing Bottom-up parsing handle Grammar Parsing, LALR parsing, parsing ambiguous grammars, YACC p specification.						programming					
	Se	mantics Syn	tax direc	ted trar	slation	S-attribu	ited and	L-attribute	d grammars		
UNIT (10 H	C-III Int Irs) ov	ermediate code atements Cont cognizers. Typ erloading of fur	e – abstrac ext Sens be checkinctions an	t syntax sitive fe ing, typ d operati	tree, trans atures – e conver ons.	slation of Choms rsions, e	simple sta ky hieran equivalenc	atements and rchy of la e of type	d control flow nguages and expressions,		

UNI	T-IV	Run time storage : Storage organization, storage allocation strategies scope access to now local names, parameters, language facilities for dynamics storage allocation.								
(10	Hrs)	Code optimization: Principal sources of optimization, optimization of basic blocks,								
		peephole optimization, flow graphs, Data flow analysis of flow graphs.								
UN	IT-V	Code generation: Machine dependent code generation, object code forms, generic code								
(14	lI-v lHrs)	$\left \begin{array}{c} \mathbf{v} \\ \mathbf{s} \end{array} \right $ generation algorithm, Register allocation and assignment. Using DAG representation of								
(1-	1113)	Block.								
Text	Books	:								
1	Introd	uction to Automata Theory, Languages and Computation, J. E. Hopcroft, R.Motwani and								
1.	J. D. l	Jllman, 3rd Edition, Pearson, 2008.								
2.	Comp	ilers Principles, Techniques and Tools Aho, Ullman, Ravisethi, Pearson Education.								
Refe	rence E	Books:								
1.	Loude	n: "Compiler Construction, Principles & Practice", 1st Edition, Thomson Press, 2006.								
2	Tremb	blay J P, Sorenson G P: "The Theory & Practice of Compiler writing", 1st Edition, BSP								
۷.	public	ation, 2010.								
3.	Theory	y of Computation, V. Kulkarni, Oxford University Press, 2013								



	Code	Category	L	Т	Р	С	I.M	E.M	Exam		
B20	AM3203	PC	3	0	0	3	30	70	3 Hrs.		
		·									
			SOFTW	ARE EN	IGINEE	RING					
(Common to AIML & CSD)											
Course Objectives:											
1.	Give exposure to phases of Software Development, common process models including Waterfall, the Unified Process, and elements of the Agile Process.										
2.	Give exposure to a variety of Software Engineering practices such as Requirements Analysis and Specification.										
3.	Give exp	osure to Software	e Design Te	echnique	s.						
4.	Give exp	osure to various S	Software Q	uality As	ssurance	and Testi	ng strateg	ies.			
				-							
Cours	e Outcom	es: At the end of	the course	Students	s will be a	able					
C N				0.4					Knowledge		
5. NO				Outcom	ie				Level		
1.	Understa	nd different softw	vare proces	s models	and their	r significa	ance.		K2		
2.	Distingui	sh various requir	em <mark>ent</mark> s ide	ntificatio	n proced	ures.			K3		
3.	Demonstr	a <mark>te d</mark> ifferent met	hods for re	quireme	nt analysi	s modelii	ng.		K3		
4.	Illustrate	various aspects o	of system de	esign and	d software	e architec	tures.		K2		
5.	Apply sof	tware quality ass	surance and	l testing	strategies	C CO			K3		
			ENQ			UUU					
		Estd. 1980		SYLLA	ABUS	IMUUS					
	The	Nature of Softw	vare, The	Unique	Nature o	f WebAp	ops, Softw	ware Engin	neering, The		
UNI	Softv	vare Process, Se	oftware Er	ngineerin	ig Practio	e, Softw	are Myth	ns. A Gen	eric Process		
(11 H	rs) Mod	el, Process A	Assessment	and	Improve	ment, I	Prescriptiv	ve Proces	ss Models,		
(Spec	Specialized Process Models, The Unified Process, Agility, Agility and the Cost of Change,									
	Agile	e Process, Extren	ne Program	nming (X	P), Other	Agile Pr	ocess Mo	dels.			
	Dage	iromonto Encia	anina T	atobl: at	ng the	Crownel		ligiting D	amiromanta		
UNIT	-II	loping Use Cas	leering, E	ng the	ng the Pequirer	Ground opts Mo	del Neg	otisting R	equirements,		
(8 Hı	(s) $\begin{bmatrix} Deve \\ Valie \end{bmatrix}$	lating Requirement	ents Requi	irements	Analysis	ients wio	uei, neg	ottating K	equitements,		
	v and	lating Requirement	ints, Requ	irements	741141 y 515	•					
	Scen	ario-Based Mo	deling. U	ML M	odels Th	nat Supp	lement 1	the Use	Case Data		
UNIT	-III Mod	eling Concepts.	Class-Base	ed Mode	eling. Re	auiremer	nts Mode	ling Strate	egies. Flow-		
(9 Hi	s) Orien	nted Modeling,	Creating a	Behavio	oral Mod	el, Patter	ms for Re	equirement	s Modeling,		
	Requ	irements Modeli	ng for Wel	o Apps.				1	Ċ,		
	<u>i</u>										
UNIT	-IV Desi	gn within the Co	ntext of So	ftware E	Ingineerir	ng, The D	esign Pro	cess, Desig	gn Concepts,		
(13 H	(rs) The	Design Model,	Software	Archited	cture, Ar	chitectura	d Genres	, Architec	tural Styles,		

		Assessing Alternative Architectural Designs, Architectural Mapping Using Data Flow,										
		Components, Designing Class-Based Components, Conducting Component-Level										
Design, Component-Level Design for Web Apps, Designing Traditional Compo												
	Component- Based Development.											
Elements of Software Quality Assurance, SQA Tasks, Goals & Metrics, Statistica												
		Software Reliability, A Strategic Approach to Software Testing, Strategic Issues, Test										
UN	IT-V	Strategies for Conventional Software, Test Strategies for Object-Oriented Software, Test										
(9F	Hrs)	Strategies for Web Apps, Validation Testing, System Testing, The Art of Debugging,										
		Software Testing Fundamentals, Internal and External Views off Testing, White-Box										
		Testing, Basis Path Testing.										
Text	t Book	s:										
1	Softv	vare Engineering: A Practitioner's approach, Roger S Pressman, 7th edition McGraw Hill										
1.	High	er Education (2009)										
2.	Softv	vare Engineering, Ian Sommerville, 9th edition. Pearson (2017)										
Refe	erence	Books:										
1.	Softv	vare Engineering, A Precise Approach, PankajJalote, Wiley India, 2010.										
2.	Softv	vare Engineering, Ugrasen Suman, Cengage (2012)										
e-Re	sourc	es:										
1.	L. https://nptel.ac.in/courses/106/105/106105182/											



Cou	urse Code	Category	L	Т	Р	С	I.M	E.M	Exam		
B2	0CD3202	PE	3			3	30	70	3Hrs		
DEEP LEARNING											
	(For CSD)										
Course Objectives:											
1	Understand	and recollect b	asic co	ncepts o	of machin	ne learnin	g				
2	Understand	concepts of de	ep feed	l forward	d networl	k mechar	isms				
3	Understand and analyze the concepts of CNN, RNN models										
4	Study the concepts of auto encoders and optimization techniques										
5	Study and analyze the different DNN architectures										
Cours	e Outcomes	At the end of	the co	urse Stu	dents wil	l be able	to				
a b									Knowledge		
S. No				Outc	ome				level		
1	Utilize the basic concept of Machine learning										
2	Apply the concepts of deep feed forward networks.K3										
3	Apply the c	oncepts of CNI	N & RI	NN mod	els	_			K3		
4	Explain and	apply optimiz	ation te	chnique	es and aut	to encode	ers.	1:00	K3		
5	Identify di	fferent DNN	mode	ls and	apply 1	that knc	wledge 1	to different	K3		
	applications			_							
			EN	SVI	LADIG	NG (.OLLI	<u>-GE</u>			
	Fund	lamontals Con	conts (ino Loor	<u>unun</u>	US				
	Histo	rical Trends in	n Deen	I earnii	ng-Machi	ine Lear	ning Basia	rs. Learning	Algorithms		
UNI	T-I Super	rvised and Uns	unervis	ed Trai	ning. Lin	ear Algel	ora for ma	chine Learni	ng. Testing.		
(12 H	Irs) Cross	s-Validation, D	imensi	onality r	reduction	, Over/U	nder-fittin	g, Hyper par	ameters and		
	valid	ation sets, Bias	, Varia	nce, Reg	gularizati	on					
	I										
TINIT	Deep	feed forwar	rd net	works:	Introduc	ction, G	radient-Ba	ased Learnir	ng, Various		
	Activ	ation Function	is, erro	r functi	ons, Diff	ferentiati	on algorit	hms, Regula	rization for		
(101)	Deep	learning, Early	v Stopp	ing, Dro	op out.						
	Conv	olutional Neu	ral Net	works A	And Seq	uence M	odeling:				
	Conv	volutional N	etworł	ks: Co	onvolutio	onal op	eration,	Motivation	Pooling,		
UNI	-III Norm	nalization,									
(10 H	Irs) Sequ	ence Modelir	ng: Ro	ecurrent	Neural	Networ	rks, Bidi	rectional RI	NNs, Deep		
	Recu	rrent	D	~		a			~		
	Netw	orks, Encoder	-Decod	er, Sequ	uence-to-	Sequenc	e Archite	ctures, the I	Long Short-		
	Term	Memory.									

UNIT (8 F	 Auto Encoders Auto encoders: under complete, regularized, stochastic, denoising, contractive, Optimization for Deep Learning 								
UNI7 (10 H	More Deep Learning Architectures & ApplicationsAlexnet, ResNet, Transfer learning, Image Segmentation, Sentiment Analysis using LSTM								
Text H	Books:								
1.	Ian Good fellow, Yoshua Bengio, Aaron Courville, "Deep Learning", MIT Press, 2016 (available at http://www.deeplearningbook.org)								
2.	Charu C Agarwal, "Neural Networks and Deep Learning", IBM T. J. Watson Research Center, International Business Machines, Springer, 2018								
Refere	ence Books:								
1.	Kevin P. Murphy, "Machine Learning: A Probabilistic Perspective", MIT Press, 2012								
2	Michael Nielsen, "Neural Networks and Deep Learning", Online book, 2016								
2.	(http://neuralnetworksanddeeplearning.com/)								
3.	Li Deng, Dong Yu, "Deep Learning: Methods and Applications", Foundations and Trends in Signal Processing, 2013.								
4.	Christopher and M. Bishop, "Pattern Recognition and Machine Learning", Springer Science Business Media, 2006.								
5.	Jason Brownlee, "Deep Learning with Python", ebook, 2016								
6.	N. D. Lewis, "Deep Learning Step by Step with Python: A Very Gentle Introduction to Deep Neural Networks for Practical Data Science, 2016.								
7.	Chris Albon, "Machine Learning with Python Cookbook-practical solutions from preprocessing to Deep learning", O'REILLY Publisher, 2018								
e-Reso	purces:								
1.	https://medium.com/nybles/create-your-first-image-recognition-classifier-using-cnn-keras- andtensorflow-backend-6eaab98d14dd								
2.	https://www.analyticsvidhya.com/blog/2017/08/10-advanced-deep-learning-architectures- datascientists/								
3.	https://www.geeksforgeeks.org/cross-validation-machine-learning/								
4.	https://www.geeksforgeeks.org/activation-functions-neural-networks/								
5.	https://towardsdatascience.com/sentiment-analysis-using-lstm-step-by-step-50d074f09948								
6.	https://medium.com/@lamiae.hana/a-step-by-step-guide-on-sentiment-analysis-with-rnn-								
7	and-Istm3a29381/e314								
/. o	https://towardsdatascience.com/common-ioss-functions-in-machine-learning-46af0ffc4d23								
ð. Web I	inte:								
1	Swayam NPTEL : Deen Learning: https://onlinecourses.pptel.ac.in/noc22_cs22/preview								
1.	Swayam M TEL. Deep Learning. https://oninfecourses.npief.ac.in/hoc22_cs22/preview								

C	Code	Category	L	Т	Р	С	I.M	E.M	Exam		
B200	CD3203	PC	3	0	0	3	30	70	3 Hrs		
		CNS (CR	YPTOGI	RAPHY	AND NE	TWORK	SECURI	ΓY)			
	(For CSD)										
Course Objectives: Students are expected to learn											
1.	Overvie	w of the compu	ter securi	ty and cl	assical en	cryption te	chniques.				
2.	Workin	g principles and	utilities	of various	s cryptogr	aphic algo	rithms inc	cluding syn	nmetric key		
	Design	issues and work	ing pring	inles of h	y algorith	uns.	est algori	thms and y	arious		
3.	authentication protocols										
4.	Various	secure commu	nication p	orotocols	standards	•					
5.	Concep	ts of firewalls a	nd block	chain tecl	hnology.						
Cours	e Outco	mes: At the end	of the co	urse stud	ents will l	be able to					
									Knowledge		
S. No				Outco	me				level		
1.	Understand Information Security goals, classical encryption techniques and acquire fundamental knowledge on the concepts related to cryptography.										
2.	Compare problem	re and apply d is <mark>related to co</mark> n	ifferent of fidentiali	encryptio ty.	n and de	cryption t	echniques	s to solve	K3		
3.	Apply perform and aut	the knowledge ance of differe pentication.	e of <mark>cry</mark> nt messa	<mark>ptogr</mark> aph ge diges	ic hash t algorith	functions ms for ver	and Illu rifying the	st <mark>rat</mark> e the e integrity	К3		
4.	Describ	e various netwo	rk securit	ty protoco	ols.				К3		
5.	Explore technol	the Importanc	e of syst	em secu	rity throu	gh firewal	lls and bl	ock chain	К3		
				SY	LLABUS						
UNIT (08 Hr	-I Cij (s) Bl	roduction to (oher Model, Sub ock Ciphers: Tr	C ryptogr ostitution caditional	aphy: Se and Tran Block Ci	ecurity A sposition ipher Stru	ttacks, Ser Technique cture, Bloc	vices & E s. ck Cipher	Mechanisn Design Pri	ns, Symmetric nciples.		
UNIT- (12Hr	UNIT-IISymmetric Key Cryptography: Data Encryption Standard (DES), Advanced EncryptionUNIT-IIStandard (AES), IDEA, Block Cipher Modes of Operations.(12Hrs)Public Key Cryptography: Principles, Public Key Cryptography Algorithms, Euler's Theorem, RSA Algorithm, Diffie-Hellman Key Exchange.										
UNIT- (12 Hr	Cryptographic Hash Functions: Application of Cryptographic Hash Functions, SHA and ſ-III MD5 Hrs) Algorithms, Message Authentication Functions, HMAC & CMAC. Digital Signatures: DSS.										

	DSS with RSA.							
	User Authentication: Remote User Authentication Principles, Kerberos							
	Electronic Mail Security: Pretty Good Privacy (PGP) And S/MIME.							
UNIT-	IP Security : IP Security Overview, IP Security Architecture, Authentication Header,							
(10 H)	Encapsulating Security Payload.							
(10 11	Transport Level Security : Web Security Requirements, Secure Socket Layer (SSL) and							
	Transport Layer Security (TLS).							
	Firewalls: Characteristics, Types of Firewalls, Placement of Firewalls, Firewall							
UNIT-	V Configuration, Trusted Systems.							
(10 Hr	Blockchain Technology: Introduction to Blockchain Technology Fundamentals, how							
	blockchain works-Shared Ledger, Permissions, Concensus, Smart contracts							
Text I	Books:							
1	Cryptography and Network Security- William Stallings, Pearson Education, 7th Edition							
2	Cryptography, Network Security and Cyber Laws – Bernard Menezes, Cengage Learning, 2010							
2	edition.							
Refer	ence Books:							
1	Cryptography and Network Security- Behrouz A Forouzan, Debdeep Mukhopadhyaya, Mc							
1.	wHill, 3rd Edition, 2015.							
2.	Network Security Illustrated, Jason Albanese and Wes Sonnenreich, MGH Publishers, 2003.							
	Computer Graphics: Principles and Practice, John F. Hughes, Andries van Dam, Morgan							
3.	Guire, David F. Sklar, James D. Foley, Steven K. Feiner, Kurt Akeley, 3rd Edition,							
	Addison- Wesley Professional, 2013.							
1	Mathematical and computer programming techniques for computer graphics, Peter Comninos,							
4.	Springer, 2010.							
5	Blockchain Fundamentals- Ravindhar vadapalli,							
5.	https://www.researchgate.net/publication/345045424_							
e-Reso	urces:							
1	https://nptel.ac.in/courses/106/105/106105031/ lecture by Dr. Debdeep Mukhopadhyay IIT							
1.	Kharagpur [Video Lecture]							
2	https://nptel.ac.in/courses/106/105/106105162/ lecture by Dr. Sourav Mukhopadhyay IIT							
۷.	Kharagpur [Video Lecture]							
2	https://www.mitel.com/articles/web-communication-cryptography-and-network-security web							
э.	articles by Mitel Power Connections.							

C	ode	Category	L	Т	Р	С	I.M	E.M	Exam		
B20A	M3206	PE	3			3	30	70	3 Hrs.		
			•	•	•	•	•	1	•		
DISTRIBUTED SYSTEMS											
(Common to AIML & CSD)											
Course Objectives: Students are expected											
1	To understand the foundations of distributed systems.										
2	To le	To learn issues related to clock Synchronization and the need for global state in distributed									
2	system	systems									
3	To lea	urn distributed n	nutual exc	lusion an	d deadloc	k detecti	on algorithr	ns			
1	To u	nderstand the	significan	ce of ag	greement,	fault to	lerance an	d recovery	protocols in		
4	Distri	buted Systems									
5	To lea	arn the character	ristics of p	beer-to-pe	er and dis	stributed	shared men	nory systems			
Course	e Outco	mes: At the end	d of the co	ourse stud	ents will	be able to)				
S No				Oute	nme				Knowledge		
5.10				Out	luine				Level		
1	Elucid	ate the foundati	ons and is	sues of d	istributed	systems			K2		
2	Illustra	ate the various	s synchro	nization	issues ar	id global	state for	distributed	K3		
	syster	ns	3)						11.5		
3	Illustra syster	nte th <mark>e Mutual</mark>	Exclusion	and Dea	dlock det	ection al	g <mark>orithms i</mark> n	distributed	К3		
4	Descri syster	be the agreements	ent protoc	ols and fa	ault toler:	ance mec	hanisms in	distributed	К3		
5	Descri	be the features	of peer-to	-peer and	distribute	ed shared	memory sy	vstems	K3		
				SY	LLABUS	5					
	Ι	Distributed Sys	stems: De	finition,	Relation	o compu	ter system	components	, Motivation,		
	F	Relation to para	allel syste	ms, Mess	sage-passi	ing syste	ms versus	shared mem	ory systems,		
	I	Primitives for d	istributed	commun	ication, S	ynchrono	ous versus a	asynchronou	s executions,		
UNI	г.i I	Design issues an	d challen	ges.							
(12 H	Irs)	a model of dis	tributed	computa	tions: A	distribute	ed program	, A model o	of distributed		
(е	xecutions, Mod	lels of cor	nmunicat	ion netwo	orks, Glo	bal state, C	uts, Past and	future cones		
	C	f an event, Moo	lels of pro	cess com	municatio	ons.					
	I	Logical Time: A framework for a system of logical clocks, Scalar time,									
	ł	hysical clock s	ynchroniz	ation: NT	Р.						
	- I -			. .			1	<u> </u>			
TINIT		lessage Order	ing & Sn	apshots:	Message	ordering	and group	communicati	on: Message		
		ordering parad	igms, A	synchron	ous exec	cution v	with synch	ronous cor	nmunication,		
	irs) [S	ynchronous pro	ogram ord al order	Global or	asynchro	mous sys	recording	o communica	ution, Causal		
	order (CO), Total order. Global state and snapshot recording algorithms: Introduction,										

		System model and definitions, Snapshot algorithms for FIFO channels.							
UNIT (10 F	F-III Irs)	Distributed Mutex & Deadlock : Distributed mutual exclusion algorithms: Introduction – Preliminaries – Lamport's algorithm – Ricart-Agrawala algorithm – Maekawa's algorithm – Suzuki–Kasami's broadcast algorithm. Deadlock detection in distributed systems: Introduction – System model – Preliminaries – Models of deadlocks – Knapp's classification – Algorithms for the single resource model, the AND model and the OR model.							
UNIT-IV (8 Hrs)		Recovery & Consensus : Check pointing and rollback recovery: Introduction – Background and definitions – Issues in failure recovery – Checkpoint-based recovery – Log-based rollback recovery – Coordinated check pointing algorithm – Algorithm for asynchronous check pointing and recovery. Consensus and agreement algorithms: Problem definition – Overview of results – Agreement in a failure – free system – Agreement in synchronous systems with failures.							
UNI' (10 F	Γ-V Irs)	Peer-to-peer computing and overlay graphs: Introduction – Data indexing and overlays – Chord – Content addressable networks – Tapestry. Distributed shared memory: Abstraction and advantages – Memory consistency models –Shared memory Mutual Exclusion.							
TEXT	ГВОО	oks:							
1.	Distr Kind	ibuted Systems Concepts and Design, George Coulouris, Jean Dollimore and Tim berg, Fifth Edition, Pearson Education, 2012.							
2.	Distr Sing	ibuted computing: Principles, algorithms, and systems, Ajay D Kshemkalyani and Mukesh hal, Cambridge University Press, 2011.							
REFF	EREN	CE BOOKS:							
1.	Distr 2007	ibuted Operating Systems: Concepts and Design, Pradeep K Sinha, Prentice Hall of India,							
2.	Adva Hill,	Advanced concepts in operating systems. Mukesh Singhal and Niranjan G. Shivaratri, McGraw- Hill, 1994.							
3.	Distr Educ	ibuted Systems: Principles and Paradigms, Tanenbaum A.S., Van Steen M., Pearson eation, 2007.							
e-Res	ource	s:							
1.	https	://nptel.ac.in/courses/106/106/106106168/							

Code		Category	L	Т	Р	С	I.M	E.M	Exam	
B20CS3207		PE	3			3	30	70	3 Hrs.	
NETWORK PROGRAMMING										
(Common to CSE & CSD)										
Course Objectives: Students are expected to										
1	Learn understanding of core network programming by using sockets and transport layer protocols like TCP and UDP									
2	Learn IPC in	earn understanding of inter process communication and implementation of different forms of C in client-server environment								
3	Get an transp	Get an exposure to various application layer protocols which are designed using sockets and transport layer protocols								
	1	<u> </u>								
Course	e Outco	mes: At the en	d of the c	ourse stu	dents wi	ll be able	to			
S. No				Outo	come				Knowledge Level	
1	Expla	in the client-ser	ver parac	ligm and	socket st	ructures			К3	
2	Describe the basic concepts of TCP sockets and TCP echo client-server K3								К3	
3	Discu	ss th <mark>e UDP soc</mark>	kets and	UDP ech	o client-s	erver prog	grams		K3	
4	Expla	in Socket option	ns and ab	ility to u	nderstand	I IPC.			K3	
5	Apply applic	the application the application of the application	ons of so , TELNE	ockets an T etc.	d demo	nstrate sk	ill to desi	ign simple	К3	
		1700		G						
		T. 4 1	NIA	5	YLLAB		114	4 1		
TINIT	тт	Introduction I	D metwo	ork Prog		g: OSI m	lodel-trans	sport layer	protocols: ICP,	
(12 H	1 - 1 [rs)	UDP and SCTP-network architecture: client-server and peer-to-peer systems, Sockets- socket Address structures: IPv4, IPv6 and Generic-value result arguments-Byte ordering functions-Byte manipulation functions-Address conversion functions								
UNI7 (10 H	ſ-II Irs)	TCP: introduct State. Elementa concurrent serv	tion to Te ary TCP ers-Close	CP-TCP sockets e function	connection – Socket n-read an	on establis -connect-l d write fur	shment and bind-listen nctions	d terminati	on TIME_WAIT rk-exec function-	
UNIT (10 H	'-III Irs)	TCP echo cl multiplexing: shutdown funct	ient ser I/O moc ion-Poll	ver prog lels-Selec function	gram-get et functi	sockname on-TCP	e and ge echo serv	etpeername ver using	e functions I/O select function-	
UNIT (08 H	C-IV [rs)	UDP: Introduc functions-UDP Socket Options	tion to U echo clie : IPv4 so	DP-diffe ent server cket optic	rence ber r program ons-IPv6	ween TC n-UDP ec socket op	P and UD ho client s tions	P-recvfron server usin	n() and sendto() g select function.	

		Socket Options: Generic socket options-TCP socket options. IPC: Introduction to IPC-								
UNI	T-V	forms of IPC-UNIX kernel support for pipes, FIFO, message queues, semaphores and								
(10	Hrs)	shared memory Network programming concepts Implementation: FTP-ping-arp-SMTP-								
		TELNET								
Textl	oooks:									
1	Unix	Network programming, the socket networking API, W.Richard Stevens, bill fenner,								
1.	Andre	Andrew m.rudoff ,PHI.								
Refe	Reference Books:									
1.	Advan	Advanced programming in the UNIX environment, W.Richard Stevens ,pearson education								



Code		Category	L	Т	Р	С	I.M	E.M	Exam		
B20CD3204		PC			3	1.5	15	35	3Hrs		
		I	I		1	1			•		
COMPUTER NETWORKS LAB											
(For CSD)											
Cour	Course Objectives: Students are expected										
1.	1. Understand and apply different network commands										
2.	2. Analyze different networking functions and features for implementing optimal solutions										
3.	3. Apply different networking concepts for implementing network solution										
4.	Impleme	nt different n	etwork pro	otocols							
Cour	se Outcor	mes: Student	ts will be al	ble to							
									Knowledge		
S. N	D			Ou	tcome				level		
1.	Implem	ent data link	layer fram	ing met	hods like	error cont	trol and f	low control.	K3		
2.	Examin	es and imple	ement the v	arious R	Routing a	lgorithms.			К3		
3.	Develop	p client-serve	er applicati	ons usin	ig sockets	s.			K3		
		10 the									
	6		181	LIST O	F PROC	FRAMS					
1.	Impleme	en <mark>t the d</mark> ata li	ink layer fr	ami <mark>ng</mark> n	nethods s	uch as cha	racter stu	<mark>if</mark> fin <mark>g an</mark> d bit s	tuffing		
2.	Write a 0	C p <mark>rogram t</mark> o	o develop a	DNS cl	ient serve	er to resolv	ve the giv	ven hostname.			
3	Impleme	ent on a data	a set of ch	aracters	the thre	e CRC pe	olynomia	ls – CRC-12,	CRC-16 and		
5.	CRC-CC	CIP 10. 1980			HUI	encenne					
4.	Impleme	ent Dijkstra's	s algorithm	to comp	oute the s	hortest pa	th in a gra	aph.			
5.	Write a G	C program to	perform s	liding w	indow pr	otocol.					
6.	Take an	example sub	onet of host	s. Obtai	n broadca	ast tree for	'it				
7.	Take an	example su	ubnet grap	h with	weights	indicating	delay b	etween nodes.	Now obtain		
	Routing	table art ead	ch node usi	ng dista	nce vecto	or routing	algorithm	1.			
8.	Write a d	client-server	applicatior	using T	ГСР.						
9.	Get the I	MAC or Phy	sical addre	ss of the	system ı	using Add	ress Reso	lution Protoco	1?		
10.	Simulate	e the Open S	Shortest Pa	th First	(OSPF) 1	routing pro	otocol ba	sed on the cos	st assigned to		
T . 4	the path.										
Textb	000KS:	4 NI - 4		- C T	1	Desili	XX 7 - 41 1				
1.	5th oditi	on 2013	s", Andrew	/ S. Tan	enbaum,	David J.	wetheral	ii, Pearson Edi	ication India;		
	"Data ($\frac{011, 2015}{Communicat}$	ion and	Networl	vina" F	Pehrouz	A Foro	uzan McGray	w Hill 5th		
2.	Edition	2017	ion and	1 Ct WOII	xing, i		A. 1010		w IIII, Jui		
3.	"Java Ne	etwork Progr	amming".	Elliotte	Rustv Ha	rold . Fou	rth Editic	on. Orielly 201	3.		
Refe	rence Boo	ks:	B ,			,104		, <u>, -</u> 01			
1	"An Intro	oduction to (Computer N	Jetwork	ing". Kei	nneth C N	Iansfield	Jr and James	L. Antonakos		
1.			- mpater 1					wild builled .	i intoinuitos		

	Pearson Education Asia.
2.	"Computer Networking, A Top-Down Approach Featuring the Internet'James F. Kuross, Keith
-	W. Ross, Third Edition, Addison Wesley, 2004.



	CodeCategoryLTPCI.ME.M								Exam	
B20)AM3209	PC	0	0	3	1.5	15	35	3 Hrs.	
					L			•		
ALGORITHMS FOR EFFICIENT CODING LAB										
(Common to AIML & CSD)										
Cou	Course Objective:									
1. To develop efficient coding for implementing advanced trees and algorithms with various inputs.										
	· · · ·									
Cou	Course Outcomes: At the end of the course Students will be able to									
S. No.									Knowledge	
5. NU				Outeo	onne				Level	
1.	Develop participation of the develop participation of the development	rograms to fi strategies.	nd optima	l solutio	ns for va	rious prot	olems us	sing different	К3	
2.	Analyze ti	me complexit	y of variou	us algori	thm desig	gn techniq	ues		K4	
3.	Develop pr	rograms to in	plement a	dvanced	trees and	l pattern m	natching	g algorithms	K3	
List	of Experim	ents:								
Impl	ement and a	nalyze the fol	llowing Al	gorithms	s using D	ivide and	Conque	r		
1.	Binary Sea	rch								
2.	Merge Sort	7 🚍 V								
3.	Quick Sort	ARA								
Impl	ement follow	ving Algorith	ims using (Greedy N	/lethod					
4.	Minimum-	cost spanning	g tree		EER		<u>OLL</u>	EGE		
5.	Single Sou	rce Shortest F	Path (Dijks	tra's)						
Impl	ement follow	ving Algorith	ims using l	Dynamic	program	ing				
6.	Optimal bin	nary search tr	rees							
7.	Traveling s	alesperson pi	roblem							
Impl	ement follow	ving Algorith	ims using I	Backtrac	king					
8.	N-Queens	problem								
9.	Graph Cold	oring problem	1							
	ement follow	wing Tree Op	erations							
10.7	ave Tree									
II. C	opiay Tiee	ving Dottorn	Motohing	Magnithe	na					
12 F	CMP Algorit	hm		Aigonnii	115.					
13 F	12. Kivit Algorithm									
13.1										
ТЕХ	TBOOKS:									
1.	Fundament	als of Com	nputer Al	gorithms	2nd e	dition by	Ellis	Horowitz. S	SartajSahni.	
Raia	sekharan. un	iversity press	. 2008			·				
2.	Advanced]	Data Structur	es – Peter	Brass, C	ambridge	Universit	y Press	, 2008		

Cour	se Code	Category	L	Т	Р	С	I.M	E.M	Exam		
B20	CD3205	PC			3	1.5	15	35	3 Hrs.		
	INTERNET OF THINGS LAB										
(For CSD)											
Course Objectives:											
1.	1. To know how to use various hardware components and Protocols in IoT applications										
2.	2. To Know how to develop various IoT applications										
Course Outcomes: At the end of the course Students will be able to											
a b									Knowledge		
S. No				Outc	ome				Level		
1.	Use sensor	rs, actuators,	Arduino a	nd Raspb	erry pi ir	ı IoT appli	cations		K3		
2.	Design and	d Develop va	rious IoT a	application	ons.				K4		
				SY	LLABUS						
1.	To interfac	ce Bluetooth	with Raspl	perry Pi/A	Arduino a	and write a	a program	to send sens	or data to		
	smart pho	ne using Blue	etooth.				_				
2.	To interfac	ce Bluetooth	with Raspl	perry Pi/	Arduino a	and write a	n program	to to turn O	N/OFF LED		
	when '1'/'	0' is received	d from sma	rt phone	using Bl	uetooth.					
3.	Applicatio	on of WiFi in	IoT Syster	ns							
4.	App desig	n for WiFi ap	oplication t	o ON/OF	FF Light		- - - - - - -				
5.	Use of var	ious network	protocols	in IoT sy	stems			-ur			
6.	Applicatio	on of 802.15.4	4 Zigbee in	IoT Sys	tems.	NUMUU					
7.	Design a s	simple IoT Sy	stem comp	orising se	ensor, Wi	reless Net	work con	nection, Data	1 Analytics		
8.	Design and	d Interface E	SP32 with	DC moto	or using I	.298 moto	r driver				
9.	Experimer	nt on connect	ivity of Ra	sberry Pi	with exi	sting syste	em compo	onents.			
Text E	Books:										
1.	Internet o Higher Ec	of Things: Arc ducation. 201	chitecture, 7	Design F	rinciples	and Appl	ications, 1	Rajkamal, M	cGraw Hill		
2	Internet o	of Things - A	Hands-on	Approac	h, Arshde	ep Bahga	and Vija	y Madisetti, I	Universities		
<u> </u>	Press, 1 st	edition, 2014									
Refere	ence Books	5 . - 41 Tu 4 - uu -	4 . f. Thin .	- A 1	. M.E.				1 1.St		
1.	2014.	g the Interne	t of Thing	gs, Adria	n McEw	en and H	akim Ca	ssimally, Wi	ley, 1 st edition,		
2.	Getting S	tarted with th	ne Internet	of Thing	s CunoPf	ister,Oreil	ly. 2011				
3.	Getting S (SPD),20	tarted with R 14.	aspberry P	i, Matt R	lichardso	n & S	Shawn W	allace, O	9;Reilly		
e-Reso	ources:										
1.	Introducti	ion to Interne	t of Things	s, https://	swayam.	gov.in/nd1	_noc20_	cs66/preview	7		
2.	An Introd	luction to Pro	ogramming	the Inter	met of Th	ings(IoT)	specializ	ation,			
² . <u>https://www.coursera.org/specializations/iot</u>											

Code		Category	L	Т	Р	С	I.M	E.M	Exam	
B20CD3206		SOC	1	0	2	2		50	3 Hrs.	
MERN STACK TECHNOLOGIES-MODULE I										
(Skill Oriented Course)										
	(For CSD)									
Cours	e Objective	• •								
1.	1. The core concepts of frontend and dynamic, responsive development for web applications.									
Cours	e Outcome	s: At the end	of the cour	rse Stude	ents will b	be able to				
~ • •									Knowledge	
S. No				Outco	ome				Level	
1.	Develop sta	atic web page	es using H	ГМL & (CSS elem	ents			K4	
2.	Develop dy	namic web p	pages and v	validate t	hem usin	g JavaScri	pt		K4	
3.	Develop I/	O Intensive V	Neb Pages	using No	odeJS				K4	
List of	f Experime	nts:								
1.	 Introduction to Web, Overview of Web Technologies, HTML - Introduction, HTML - Need, Platform-independency, DOCTYPE Declaration, Types of Elements, HTML Elements – Attributes, Paragraph Element, Division and Span Elements, List Element, Link Element, Character Entities, HTML5 Global Attributes, Creating Table Elements, Table Elements : Colspan/ Rowspan Attributes, border, cellspacing and cellpadding attributes, Creating Form Elements, Input Elements - Attributes, Color and Date Pickers, Select and Data list Elements, Editing Elements, Media. CSS: Introduction CSS, Applying CSS to HTML, Selectors, Properties and Values, CSS Colors and Backgrounds, CSS Box Model, CSS Margins, Padding, and Borders, CSS Text, and Fort 								TML - Need, Elements – Link Element, le Elements : Creating Form list Elements, SS Colors and Cext and Font	
	JAVASCI	RIPT:								
2.	JAVASCRIPT: Why we need JavaScript, What is JavaScript, Environment Setup, Working with Identifiers, Type of Identifiers, Primitive and Non Primitive Data Types, Operators and Types of Operators, Types of Statements, Non - Conditional Statements, Types of Conditional Statements, If and Switch Statements, Types of Loops, Types of Functions, Declaring and Invoking Function, Arrow Function, Function Parameters, Nested Function, Built-in Functions, Variable Scope in Functions, Working With Classes, Creating and Inheriting Classes, In-built Events and Handlers, Working with Objects, Types of Objects, Creating Objects, Combining and cloning Objects using Spread operator, Destructuring Objects, Browser and Document Object Model, Creating Arrays, Array Methods.									
3.	NodeJS: What is N Exports	odeJs, Funct	ions, Buffe	er, Modu	ıles & Ty	pes, Core	Modules	, Local Mod	ules, Modules	

	What is NPM?, Installing Packages Locally, Installing package globally, Adding dependency in							
	package Json, Updating packages							
	Creating Web Server, Sending Requests, Handling HTTP requests,							
	File System - Read File, Writing a File, Opening a File, Deleting a File, Writing a file							
	Debugging Nede IS Application Const Nede IS Debuggen							
	Debugging Node JS Application Core Node JS Debugger							
	ExpressJ8:							
4.	Event Emitter class, Inheriting Events, Returning event emitter							
	Express JS, Configuring Routes, Working with Express							
	Serving Static Resources, Serving Static Files, Working with Middle Ware							
	TypeScript:							
	TypeScript Overview, Intro to TypeScript, Tooling and Environment, Creating a TypeScript							
5.	Project							
	TypeScript, Type System, Enums, Functions, Interfaces and Classes, Modules, Generics, Mapped							
	Types, Conditional Types, Decorators, Type Definitions, Configuration							
Text	Books:							
1.	Programming the World Wide Web, 8th Edition Robet W Sebesta, Pearson, 2015.							
2.	WebTechnologies,1st Edition 7th impression, Uttam K Roy, Oxford, 2012.							
2	Beginning MERN Stack: Build and Deploy a Full Stack MongoDB, Express, React, Node.js App,							
5.	by Greg Lim, 2021							
Refe	rences:							
1.	https://www.javatpoint.com/mern-stack							
2.	https://blog.logrocket.com/mern-stack-tutorial/ 170MOMOUS							

Code			Category	L	Т	Р	С	I.M	E.M	Exam
B20MC3201		201	MC	3						3 Hrs.
EMPLOYABILITY SKILLS										
(Common to AIDS, AIML, CSBS, CSD, CSE, ECE & IT)										
	Part-A: Verbal Ability									
Co	Course Objectives:									
1	To introduce concepts required in framing grammatically correct sentences and identifying									
errors While using Standard English.										
2.	To f	famili	arize the lear	ner wit	h high	frequen	cy words	as they	would be	used in their
	profe	ession	al career.							
3.	To in	nculca	ate logical thin	king in	order to	o frame a	nd use da	ta as per	the requirem	nent
4.	То а	cquai	nt the learner	of mal	king a c	oherent	and cohe	sive sen	tences and p	paragraphs for
	com	posing	g a written dis	course.	4 -1 '11		:	(1 •	£ ' 1	
5.	10 f	amilia	trize students v	with sof	t skills a	and how	it influen	ces their	protessional	grow.
0					1 11					
Co	urse O	utcon	nes: The stude	nts will	be able	to				17 1 1
	S.No	1	Contraction of the second		0	utcome				Knowledge Level
		Dete	ect grammatics	al errors	s in the	text/sen	tences an	d rectify	them while	•
	1	ansv	vering their	compe	etitive/co	ompany	specific	tests	and frame	K3
		gran	matically Cor	rect ser	itences	while wr	iting.	COL	LEGE	
	2	Ansv Exer	cises while at	on syn temptin	onyms, g CAT,	GRE, G	ATE and	other rel	ated tests.	K3
	3	Use	their logical th	ninking	ability	and solv	e question	ns related	d to analogy,	K3
	5	Sylle	K5							
	4	Cho	ose the approp	priate w	vord/s/pl	hrases su	itable to	the give	en context in	K3
	_	orde								
 					07	7	10			
T 75 7	TTTT	G		1 4	51		5			
UN	11-1	Spo	otting Errors, S	Sentence	e Impro	vement				
 		C	onume Anto	numa	Fragues	the Cont	Fund W	rda Ear	nian Dhraca	diama and
UN	IT-II	Syl Dhr	ionyms, Anto wasal Verba Co	nyms, 1 Mocatic	rrequen	try Con	used wo	ords, For	eign Phrase	s, laioms and
		F III		mocan	ліs. 					
TINI	т.ш	For	eion Phrases	Idiome	and Phr	asal Verl	s Colloc	ations A	Analogies O	dd One Out
		101	-1511 I III 4303,			ubui volt	, conoc	auono, r	inui02108, O	
UNI	T-IV	Ser	itence complet	ion Sei	ntence F	Equivaler	nce. Close	e Test		
		501				-941 / 1101		. 1001		
UN	IT-V	Res	ding Compres	nension	Para I	ımhles				
			compton	101101011	, 1 111 1	*1110105				

Text Bool	ζS:						
1.	Oxford Learners,, Grammar–Finder by John Eastwood, Oxford Publication.						
2.	RS Agarwal books on objective English and verbal reasoning						
3.	English Vocabulary in Use-Advanced, Cambridge University Press						
4.	Collocations In Use, Cambridge University Press						
5	Soft Skills & Employability Skills by Samina Pillai and Agna Fernandez, Cambridge						
5.	University Press India Pvt .Ltd.						
6.	Soft Skills, by Dr.K.Alex, S. Chand & Company Ltd., New Delhi						
Reference	Books:						
1.	English Grammar in Use by Raymond Murphy, CUP						
2.	Websites: Indiabix,800score, official CAT, GRE and GMAT sites						
3.	Material from IMS, Career Launcher and Time institutes for competitive exams						
4.	The Art of Public Speaking by Dale Carnegie						
5.	The Leader in You by Dale Carnegie						
6.	Emotional Intelligence by Daniel Golman						
7.	Stay Hungry Stay Foolish by Rashmi Bansal						
8.	I have a Dream by Rashmi Bansal.						
	Part-B: Quantitative						
Aptitude-I							
Course Objectives: ENGINEERING COLLEGE							
1.	To familiarize students with basic problems on numbers and ratios problem.	ems.					
2.	To enrich the skills of solving problems on time, work, speed, distance and also						
	To enable the students to work efficiently on percentage values related to shares						
3.	profit and Loss problems.						
4.	To inculcate logical thinking by exposing the students to reasoning related questions						
5.	To inculcate logical thinking by exposing the students to reasoning related questions.						
Course O	utcomes: The students will be able to						
S.No.	Course Outcome	Knowledge Level					
1	The students will be able to perform well in calculating on number	K3					
1.	problems and various units of ratio concepts	K5					
2.	The students will be able to solve problems on time and distance and units related solutions K3						
3.	The students will become adept in solving problems related to profit and loss, in specific, quantitative ability K3						
4.	The students will present themselves well in the recruitment process using analytical and logical skills which he or she developed during K3						
	and the region of the of the developed during						

	the course as they are very important for any person to be placed in the industry						
	The students will some to apply Logical thinking to the puckleme of						
_	The students will earn to apply Logical thinking to the problems of						
5.	Syllogisms and be able to effectively attempt competitive	K3					
	examinations like CAT, GRE, GATE for further studies						
	SYLLABUS						
	Numbers, LCM and HCF, Chain Rule, Ratio and Proportion Importance	of different					
	types of numbers and uses of them: Divisibility tests, finding remainders in various						
LINITT I	cases, Problems related to numbers, Methods to find LCM, Methods to find HCF,						
0111-1	applications of LCM, HCF. Importance of chain rule, Problems on chain rule,						
	Introducing the concept of ratio in three Different methods, Problems related to Ratio						
	and Proportion						
I	*						
	Time and work, Time and Distance Problems on manpower and time rela	ted to work,					
	Problems on alternate days, Problems on hours of working related to clock, Problems						
	on pipes and cistern, Problems on combination of the some or all	the above,					
UN11-11	Introduction of time and distance, Problems on average speed, Problems on Relative						
	speed, Problems on trains, Problems on boats and streams, Problems on circular						
	tracks, Problems on polygonal tracks, Problems on races.						
duras, riotenis on polygonal duras, riotenis on fues.							
	Percentages, Profit Loss and Discount, Simple interest, Compou	nd Interest,					
	Partnerships, shares and dividends. Problems on percentages-Understan	ding of cost					
	price selling price marked price discount percentage of profit percentage of loss						
UNIT-III	percentage of discount. Problems on cost price selling price, market price discount						
	Introduction of simple interest Introduction of compound interest Relation between						
	simple interest and compound interest. Introduction of partnership. Sleeping partner						
	concept and problems. Problems on shares and dividends, and stocks						
I	concept and problems, ricoronis on shares and arriands, and scorist						
	Introduction number series number analogy classification. Letter ser	ies, ranking					
	directions Problems of how to find the next number in the series. Finding the missing						
	number and related sums. Analogy Sums related to number analogy Ranking of						
UNIT-IV	alphabet Sums related to Classification Sums related to letter series Relation						
	between number series and letter series. Usage of directions north south east west						
	Problems related to directions north south east west						
	Toblems related to directions norm, south, east, west.						
 	Data sufficiency. Syllogisms Easy sums to understand data sufficience	w Frequent					
UNIT-V	mistakes while doing data sufficiency. Syllogisms Problems	ly, Mequein					
	mistakes while doing data sufficiency, Syfiogisilis Flobicilis.						
Text Book	Text Books:						
1.	Ouantitative aptitude by RS Agarwal						
2.	Verbal and nonverbal reasoning by RS Agarwal						

3.	Puzzles to puzzle you by shakunatala devi.				
References:					
1.	Barrons by Sharon Welner Green and IraK Wolf (Galgotia Publications pvt. Ltd.)				
2.	Websites: m4maths, Indiabix, 800score, official CAT, GRE and GMAT sites				
3.	Material from IMS, Career Launcher and Time,, institutes for competitive exams				
4.	Books for CAT by Arun sharma.				
5.	Elementary and Higher algebra by HS Hall and SR Knight.				
Websites:					
1.	www.m4maths.com				
2.	www.Indiabix.com				
3.	www.800score.com				
4.	Official GRE site				
5.	Official GMAT site				



Code		Category	L	Т	Р	С	I.M	E.M	Exam
B20HS3204		HS	2						
			GEN	NDER S	SENSITI	ZATION			
			(Co	ommon t	o ALL B	ranches)			
Cour	se Obje	ectives:							
1.	To dev	To develop students' sensibility with regard to issues of gender in contemporary India.							
2.	To pro	To provide a critical perspective on the socialization of men and women.							
3.	To intr	Γο introduce students to information about some key biological aspects of genders.							
4.	To hel	To help students reflect critically on gender violence and workplace security.							
5.	To expose students to more egalitarian interactions between men and women.								
Cour	se Outo	comes: At the er	nd of the co	ourse, st	udents w	ill be able	to		
S.No				Outco	ome				Knowledge
0.110				oute	,				Level
1.	Under	stand the impor	tant issues	relating	to gende	er in conte	mporary	India.	K2
2.	Get	sensitized to	basic dir	nension	s of th	ne biolog	gical, so	ociological,	К2
	psych	ological and leg	al aspects of	of gende	er.				
3.	Attain a finer grasp of how gender discrimination works in our society and						K2		
	how to	o counter it.							
4.	Acqui	re insight into tl	ne gendere	d divisio	on of lab	our and it	s relatior	to politics	K2
	and ed	conomics.	<u> </u>					1 0110	
5.	Devel	op a sense of ap	preciation	for both	men and	l women 1	n all wal	ks of life.	K3
		Estd. 1980				NOMO	<u>US</u>		
			<u>C</u> J	SY.)		FJT	•e-
Understanding Gender and Related Concepts - Gender in Everyday Life						lle			
LINI	T_T	Introduction: Conceptual Connotation – Sex and Gender – Basic Gender Concepts -							
UI		Gender Poles & Polationships Mutha Conder in Indian acciety Early days Later							
		Vedic Period Medieval and British Period Independent India							
			icult vul ul			macpen	aont mu		
	[]	Introduction to	Gender J	ustice-]	Notion a	nd Signif	icance		
		Division and Valuation of Work – Housework- The Invisible Work - "My Mother							
UNI	T-II	I doesn't work," - Offences against Women –Fact and Fiction - Status of Women in							
		Society – Gender and Human Rights - Gender Equality – Gender Justice – Notion and							
		Significance							
]	International a	nd Constit	utional	Perspec	tives on (Gender H	Equality	
		The International Bill of Rights, 1979 –Declaration on the Elimination of Violence							
UNI	L-III 8	against women 1993 – The Rights of Women – Beijing Platform for Action 1995 –							
	(Constitutional Guarantees – Fundamental Rights – Equality.							
	I								

		Gender and Culture			
UNIT	T T T 7	Gender and Film - Gender and Electronic Media - Gender and Advertisement - Gender			
	L - I V	and Popular Literature – Gender Issues - Gender-Sensitive Behaviour – Gender being			
		Together as Equals.			
		Gender Violence- Within and Beyond			
TINIT	T X7	Violence – Gender Violence – Types of Gender Violence –Gender Violence in Indian			
UNL	Perspective – -Women Specific Legislations for the Elimination of Violence Within a				
		Beyond.			
Refer	ence	Books:			
"Towards A World Of Equals: A Bilingual Textbook on Gender" by A. Suneetha, U					
1.	Bhru	ugubanda, Duggirala Vasanta, Rama Melkote, Vasudha Nagaraj, Asma Rasheed, Gogu			
	Shy	amala, Deepa Sreenivas, and Susie Tharu, Published by Telugu Akademi (2015).			
_	Ferb	er, Holcomb & Wentling, Sex, Gender & Sexuality: The New Basics, Oxford Univ.			
2.	Pres	s 2008.			
2	Flav	ia Agnes, Sudhir Chandra, Monmayee Basu, Women and Law in India, Oxford Univ.			
3.	Pres	s 2004.			
4.	Mamta Rao, Law Relating to Women and Children, Eastern Book Co, Lucknow.				
5.	K.I. Vibhute, Criminal Law, Lexis Nexis, 12th Edn.				
6.	N. P	rabha Unnithan (ed.), Crime & Justice in India, Sage Pub., 2013.			
7.	Ritu Gupta, Sexual Harassment at Workplace, Lexis Nexis, 2013.				
0	IGN	OU: Gender Sensitization: Society, Culture and Change (2019) BGSE001, New Delhi			
0.	IGN	OU.			
	•	ENGINEERING COLLEGE			
Web	links	Estd. 1980 AUTONOMOUS			
1.	http	s://nptel.ac.in/courses/110105080			
2.	https://www.youtube.com/watch?v=2Xfp2eiTte0				
3.	https://www.youtube.com/watch?v=-FCEBe5VNcA&t=41s				
4.	https://www.youtube.com/watch?v=7n9IOH0NvyY				
5.	https://www.youtube.com/watch?v=dpC2jGqu4G0				
6.	https://www.youtube.com/watch?v=kcW4ABcY3zI&t=99s				
7.	https://www.youtube.com/watch?v=dlXw1PbnWKM				
8.	https://www.youtube.com/watch?v=9bayaZ18_po				
9.	https://www.youtube.com/watch?v=ZbLq23cGFV4&t=1662s				
10.	https://www.youtube.com/watch?v=61aYvb0Vo68				
11.	https://www.youtube.com/watch?v=728H4Khf7Gk&t=1793s				
12.	https://www.youtube.com/watch?v=y2Yk-rSZ7PI				
13.	https://www.youtube.com/watch?v=wSqFvcjDpos				
14.	https://www.youtube.com/watch?v=AljDd7nj9wE				
15.	https://www.youtube.com/watch?v=MKPM0f2fOjM				