# SEMESTER V

S. No	Category	Course Code	Course Title	L	T	P	C
Theory							V 1
1	PC	CS23501	Theory of Computation	3	1	0	4
2	PC	CS23502	Data Warehousing and Data Mining	3	0	0	3
3	PC	CS23503	Artificial Intelligence	3	0	0	3
4	PC	CS23504	Software Engineering	3	0	0	3
5	PC	CS23505	Foundation of Data science	3	0	0	3
6	PE	CS2315*	Professional Elective – I	3	- 0	0	3
Practi	cal						211
7	PC	CS23506	Data Warehousing and Data Mining laboratory	0	0	2	1
8	PC	CS23507	Artificial Intelligence laboratory	0	0	4	2
9	EE	EE23502	Industrial Training *	0	0	2	1
10	EE	GE23501	Professional Development III	0	0	2	1
10	1 22		Total	18	1	10	24

# Two weeks during Summer Vacation

# SEMESTER VI

S. No	Category	Course Code	Course Title	L	T	P	C
Theory		The state of		ated a service	100		1 2
1	. HS	BA23***	Entrepreneurship Development	3	0	0	3
2	PC	CS23601	Cryptography and Network Security	- 3	0	0	3
3	PC	CS23602	Big Data Analytics	3	0	0	3
4	PC	CS23603	Machine Learning	3	0	0	3
5	PE	CS2325*	Professional Elective – II	3	0	0	3
6	OE	CS2390*	Open Elective – I	3	0	0	3
Practic				11.			
7	PC	CS23604	Big Data Analytics laboratory	0	0	4	2
8	PC	CS23605	Machine Learning Laboratory	0	0	4	2
9	EE	EE23601	Design Thinking	0	0	2	1
Talker by			Total	18	0	10	23



CS3	3501	THEORY OF COMPUTATIO	N		3	0	0	3
CO	URSE O	BJECTIVES			,			
То	enable the	e students to				2 4 4		
1	underst	and the need for automata theory in computation	1.	2 8			gir il- s	
2	analyze	and construct finite automata and regular expre	ssions for re	egular l	angi	ıage	s.	Final Co.
3	develop	and simplify context-free grammars and analyz	e their amb	iguity.			1. 5.	
4	design	pushdown automata and establish equivalence w	ith CFGs.	1		50		- 3, - 3
5	underst	and the Turing machine model and the fundamen	ntals of con	putation	onal	com	plex	ity.
UNI	TI	INTRODUCTION TO AUTOMATA		1000			5 1	9
Nee	d for au	tomata theory - Introduction to formal proof - I	Finite Autor	nata (F	(A)	- De	term	inistic
Fini	te Autor	nata (DFA) – Non-deterministic Finite Automata	a (NFA) – I	Equival	ence	bet	weer	a NFA
		Conversion from NFA to DFA.						
71	ТПТ	REGULAR EXPRESSIONS AND LANGUAGE	<u>'S</u>					9
		ression – Regular Languages- Equivalence of Fir		ta and	rami	lore	vnre	
					e 7.5		-	• 0
		nguages to be not regular (Pumping Lemma) – Cl	losure prope	rues o	reg	uiar	lang	
_	T III	CONTEXT FREE GARMMAR						9
		e Grammar (CFG) – Definitions – Derivatio					17	
		nd languages - Chomsky Normal Form (CNF) -	- Pumping	lemma	for	CFL	. – C	losure
		f Context Free Languages			1			
	T IV	PUSH DOWN AUTOMATA						9
Pus	h Down	Automata (PDA): Definition - Moves - Insta-	ntaneous de	escripti	ons	-Lar	ngua	ges of
pus!	hdown a	utomata - Equivalence of pushdown automata a	nd CFG-Co	nversi	on of	CF	G to	PDA-
PD	A to CFC	G – Deterministic Pushdown Automata						
UN	TV	TURING MACHINES AND UNDECIDABILIT	<u> </u>					9
Tur	ing Mac	hine: Basic model – Language acceptance by	TM – Uni	versal	Tur	ing ]	Mac	hine –
Uns	solvable	Problems – Post Correspondence Problem(PCP)	– Tractable	e and I	ntrac	table	e pro	blems
		completeness – 3-CNF SAT problems.	m il				- P	
				TOTA	I. PF	RIO	DS	45
CO	URSE O	UTCOMES	produced and	JAM				13
		f the course, the students will be able to	· Stell . J	BT M				
CO	1 cons	truct finite automata and explain their significance		(Highe	st Le	vel)	, C 4	
		atomata theory.	τ	Indersta	andin	g (K	2)	

CO2	write regular expressions and apply techniques to identify non-regular languages.	Analyzing (K4)
CO3	design and simplify context-free grammars and analyze ambiguity in language definitions	Applying (K3)
CO4	design pushdown automata and demonstrate equivalence with context-free grammars.	Applying (K3)
CO5	understand the Turing machine model and analyze decidability and complexity concepts.	Analyzing (K4)

- 1. J.E.Hopcroft, R.Motwani and J.D Ullman, "Introduction to Automata Theory, Languages and Computations", Third Edition, Pearson Education, 2006.
- 2. John C Martin, "Introduction to Languages and the Theory of Computation", 4th Edition, Tata McGraw Hill, 2011.

# REFERENCES

- 1. Harry R Lewis and Christos H Papadimitriou, "Elements of the Theory of Computation", 2nd Edition, Prentice Hall of India, 2015.
- 2. Peter Linz, "An Introduction to Formal Language and Automata", 6th Edition, Jones & Bartlett, 2016.
- 3. K.L.P.Mishra and N.Chandrasekaran, "Theory of Computer Science: Automata Languages and Computation", 3rd Edition, Prentice Hall of India, 2006.
- 4. Michael Sipser, "Introduction to the Theory of Computation", 3rd edition, Cengage Publications.

# **CO-POMAPPING:**

Mapping of Course Outcomes with Programme Outcomes: (1,2,3 indicates the strength of correlation) 3 – Strong, 2 – Medium, 1 – Weak

COs					Progr	amme	Outco	mes (P	Os)				Spe Outo	ramme cific comes (Os)
	PO 1	PO 2	PO 3	PO 4	PO 5	PO 6	PO 7	PO 8	PO 9	PO1	PO1	PO12	PSO 1	PSO 2
CO1	3	2	2	-	2	-	( <b>-</b> 1)	- "	-		·	17	3	2
CO2	3	3	3	1	-	-	7 <b>-</b> k≪.	1		1-17	- 1	- 11 - N	3	3
CO3	3	2	3	- 20	-		( ):	/ <b>-</b> //	• •	- 4		Y-	3	2
CO4	3	3	3	-	V. <b>-</b> 20	-	En-	1 2 1.5 pt	1.0 <u>¥</u>	-	· . · . · .	, <b>-</b>	3	3
CO5	3	3	2	2		-:	<b>-</b>	11 3-2	-	1-		74 - 15 A	3	3



To enable  1. U 2. To 3. A 4. St 5. Le UNIT I  Data Wa modeling index and computat  UNIT II  Introducti confluence Cleaning  UNIT III  Basic con Item set m Pattern min UNIT IV  Basic Con	Unders  Unders  Fo kno  Analyz  Study v  Learn c  Interpretation mation mation mation for the learn of the l	students to  rstand the fundamental concepts of Data Warehousing and OLAP technologies  now about the basics of data mining and application of datamining techniques  rze and explore association rule mining techniques  rvarious classification techniques in datamining.  different clustering methods and advanced clustering techniques  DATA WAREHOUSE AND OLAP TECHNOLOY  ouse- Basic concepts of data warehousing, data warehouse Architecture; Data warehouse hema and measures; OLAP operations-Typical OLAP operations, Indexing OLAP data: bitman in index, Storage implementation: column-based databases; Data cube computation; Data cube methods.  DATA MINING AND PREPROCESSING  - Data Mining, Diversity of data types for data mining, Mining various kinds of knowledge of multiple disciplines, Application; Data types; Data Preprocessing- Data quality, Data ata Integration; Data Transformation; Dimensionality reduction.  PATTERN MINING  ots - Market Basket Analysis, Frequent Itemset, Closed Itemset, and Association Rules; Frequent Market Basket Analysis, Frequent Itemset, Closed Itemset, and Association Rules; Frequent Market Basket Analysis, Frequent Itemset, Closed Itemset, and Association Rules; Frequent Market Basket Analysis, Frequent Itemset, Closed Itemset, and Association Rules; Frequent Market Basket Analysis, Frequent Itemset, Closed Itemset, and Association Rules; Frequent Market Basket Analysis, Frequent Itemset, Closed Itemset, and Association Rules; Frequent Market Basket Analysis, Frequent Itemset, Closed Itemset, and Association Rules; Frequent Market Basket Analysis, Frequent Itemset, Closed Itemset, and Association Rules; Frequent Market Basket Analysis, Frequent Itemset, Closed Itemset, and Association Rules; Frequent Market Basket Analysis, Frequent Itemset, Closed Itemset, and Association Rules; Frequent Itemset, Closed Itemset, and Associ	y,
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Pattern min UNIT IV Basic Con		is methods, I attern Evaluation Methods, kinds of patterns: Constraint-based nattern mit	ung
Basic Con		g- application examples.	
	7 (	CLASSIFICATION	- 5
Classificat	Bayesi	ts-Decision tree Induction, Bayes Classification Methods, Lazy learners; Advanced Classification Belief Networks, Support vector machines, Rule-based and pattern-based classification with weak supervision Other.	tion
UNIT V		CLUSTERING AND DATA MINING APPLICATIONS	9
Detection-		sis; Partitioning methods; Hierarchical methods; Density-based and grid-based methods; Outstering-vs. classification-based approaches; Data mining trends and research frontiers-Mining tamining applications.	

COU	RSE OUTCOMES	
At the	end of this course, students will be able to	BT Mapped (Highest Level)
CO1	Understand the architecture, components, and models of data warehouses and OLAP technologies.	Understand (K2)
CO2	Apply preprocessing techniques to prepare raw data for mining	Apply (K3)
CO3	Analyze various association rule mining methods and discover patterns in large datasets.	Analyze (K4)
CO4	Compare and apply classification techniques to categorize data effectively.	Analyze (K4)
CO5	Evaluate clustering methods and interpret the application of data mining techniques in real-world scenarios.	

- 1. Jiawei Han, Jian Pei, Hanghang Tong, "Data Mining: Concepts and Techniques", 4th Edition, Morgan Kaufmann, 2023.
- 2. Alex Berson and Stephen J.Smith, "Data Warehousing, Data Mining & OLAP", Tata McGraw Hill Edition, 35th Reprint 2016.

#### REFERENCES

- Ralph Kimball & Margy Ross," The Data Warehouse Toolkit: The Definitive Guide to Dimensional Modeling, Wiley, 2013.
- 2. Lawrence Corr & Jim Stagnitto, "Agile Data Warehouse Design: A Practical Guide for Building Dimensional Data Warehouses", Technics Publications, 2017.
- 3. Pang-Ning Tan, Michael Steinbach, Anuj Karpatne & Vipin Kumar"Introduction to Data Data Mining,2017.
- 4. David Hand, Heikki Manila and Padhraic Symth, "Principles of Data Mining", PHI, 2004.

# **CO-PO MAPPING:**

Mapping of Course Outcome (CO's) with Programme Outcomes (PO's) and Programme Specific Outcomes PSO's (1/2/3 indicates strength of correlation) 3-Strong, 2-Medium, 1-Weak

					Progr	amme	Outcon	nes PO	's				P	SO's
CO's	. 1	2	3	4	5	6	7	8	9	10	11	12	1	2
CO1	3	2	2	1 - 1	2 ·	-			1 -	1	<u> </u>	2	3	2
CO2	2	3	2	7. E.	3	2	Life in the delign		-		-	2	3	3
CO3	3	3	3	2	3	-		1	-	2	1	3	3	3
CO4	3	3	3	2	3	-	÷ ;	- 10	2	2	2	3	3	3
CO5	2	3	2	3	3	1	1 b	2	2	2	2	3	3	3

Approved

Approved

BOARD OF STUDIES

Computer Science & Engineering

CS23	503		Υ .	ARTIFIC	CIALIN	TELLI	GENCE				3	0	0	3
				(Comr	non to C	CSE, IT,	CSE(IoT	))						18
COU	RSE OE	BJECTIVES												
To en	able the	students to												
1	Acqu	ire a knowledge o	of vari	ous metho	ods of di	fferent p	roblem s	olving a	and sear	ching t	techn	iques		
2	under	stand the concepts	ts of k	nowledge	represer	ntation								
3	under	stand about knov	wledg	e inferen	ice and	how to	solve th	e prob	lems us	ing va	arious	sinfe	rence	
4	realiz	e the concepts of p	plann	ing and ex	xpert sys	stem	4					11 11	. 4	
5	under	stand the method	d of va	rious AI A	Applicati	ions								
UNIT	I	INTRODUCTI	TOOI	'O AI			1 1 1				1	21		9
	stics Sea	ods - Issues in t rch – Search Tech KNOWLEDGV	hnique	es			- Intell	igent A	Agents –	- Sear	ching	g Stra	tegie	
UNIT Predict Repre UNIT Infere	eate Log senting lance in Fring – Re	KNOWLEDGV gic — Representing knowledge using F KNOWLEDGE irst Order Logic a esolutions — Rease	WE R ing kn Frame E INF and Pr	ESPRES owledge es – Script ERENCE	ENTAT using reasts – Know	TION rules — S wledge E	Symbolic Engineeri	Reasong in Fi	ning – rst Orde - Forwar	Statis r Logi rd Cha	tical c	Reas	oning	9 g –
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Heuris UNIT Predict Repre UNIT Infere Chain Dynar UNIT Comp Expert System	stics Sea III cate Log senting I III nce in Fi ing – Re nic Baye IV onents o t System n Applie	knowLEDGV gic — Representing knowledge using I KNOWLEDGE irst Order Logic a esolutions — Rease esian Networks	WE R ing kn Frame E INF and Pr soning ND E stem —	ESPRES owledge es – Script ERENCE oportiona : Probabi  XPERT S Hierarch ert System	ENTAT using rates – Know E Il Logic: listic Re EYSTEM ical Plar m – Rol	TION rules — S wledge E Unificat easoning Inning — I	Symbolic Engineeri tion and I – Bayes Reactive	Reasong in Fi	rst Orde - Forwar rks – H	Statis r Logi rd Cha idden er Plar	tical c  ining Mark	Reas  g – Ba  cov M  Tech  ition	oniną ockwa Iodel uniqu – Ex	9 33 9 ard s - 9 ees - pert
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Heuris UNIT Predict Repreduction UNIT Infereduction Chain Dynam UNIT Comp Expert System Optim UNIT Langu	stics Sea FII cate Log senting le FIII nce in Fi ing – Re nic Baye IV onents o t System n Applic izer V lage Mod	knowLEDGV cic — Representing KnowLedge using I KNOWLEDGE irst Order Logic a esolutions — Reason Networks PLANNING AN of a Planning System: Architecture of cations: Medical	WE R ing kn Frame E INF and Pr soning  ND E  of Exp Diagn TIONS	ESPRES owledge es – Script ERENCE oportiona :: Probabi  XPERT S Hierarch ert System nosis Sys	ENTAT using rests – Known E Il Logic: listic Res EYSTEM ical Plan m – Role items –	Unificate asoning  Inning — Item of Extended	Symbolic Engineeri tion and I – Bayes Reactive kpert Sys	Reasong in Fi	rst Orde - Forwar rks – H s – Othe Knowled	Statis r Logi rd Cha idden er Plan dge A	tical c tining Mark mning cquis	Reas g – Ba cov M g Tech ition ring I	oning ckwa dodel nniqu - Ex	9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9



	end of this course, students will be able to	BT Mapped (Highest Level)
CO1	demonstrate awareness of intelligent agents and problem solving using search methods	Applying (K3)
CO2	develop and analyze the knowledge about the usage of predicate logic and first order logic for making inferences	Analyzing (K4)
CO3	use the knowledge and the process of inference to derive new facts	Analyzing (K4)
CO4	Get the development ideas of planning system and expert system by using applications	Applying (K3)
CO5	apply knowledge of language models and know the ideas about the robot	Applying (K3)

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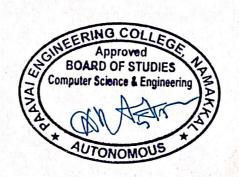
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#### **CO-PO MAPPING:**

Mapping of Course Outcome (CO's) with Programme Outcomes (PO's) and Programme Specific Outcomes PSO's

(1/2/3 indicates strength of correlation) 3-Strong, 2-Medium, 1-Weak

1.15	440	141	Marie Control	4.14 A.15	Progra	amme O	utcomes	PO's	5 - 17				PS	O's
CO's	1	2	3	4	5	6	7	8	9	10	11	12	1 1	2
CO1	3	3	3	3	2	2	-	3	- *	5 20	`\ \ <b>-</b> \)	3	3	3
CO2	3	3	3	3	2	2	-	3	-	-	-	3	3	3
CO3	3	3	3	3	2	2	1 - 1 - 1 - 1 - 1 - 1 - 1 - 1 - 1 - 1 -	3	-,.	-10	- 7	3	3	3
CO4	3	3	3	3	2	3	(	3	-	-	7-	3	3	3
CO5	3	3	3	3	2	3		3	A 1	-		3	3	3



COUR	SOFTWARE ENGINEERING RSE OBJECTIVES	3	0	0 3
	able the students to			
1	Introduce the fundamental concepts of software engineering and software	process models	s.	
2	Apply the methods for modeling software requirements through scenario on web apps.	s, classes, and be	ehaviors	s, focusi
3	Analyze the software design principles, architectural styles, and data flow architecture.	v mapping for ef	fective	software
4	Apply software testing strategies, methodologies, and techniques for conveb applications.	ventional, object	t-oriente	ed, and
5	Analyze the key concepts of software project management, metrics, and e	estimation.		
UNIT I	I INTRODUCTION TO SOFTWARE ENGINEERING	y and the same of		9
Softwar	are and Software Engineering - The Nature of software, The Unique	N-4 0 *		9
1100055	sering, The Software Process, Software Engineering Practice, Software Mys Model, Process Assessment and Improvement, Prescriptive Process Models, Process, Personal and Team Process Models, Process Technology, Process Technolog	dels. Specialized	d Proce	ss Mode
	QUALIMENTS MODELING			9
Modelli	ements Modeling: Scenarios, Information and Analysis Classes - Requiring, UML Models That Supplement the Use Case, Data Modeling C	oncents Class	D1	1 1 1 1
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Requirent Modeling UNIT II Design (	ements Modeling: Flow, Behavior, Patterns, WebApps – Requirements Mong, Creating a Behavioral Model, Patterns for Requirements Modeling, Requirements Modeling Concepts – Design within the Context of Software Engineering. The Design Within the Context of Software Engineering.	oncepts, Class- odeling Strategic irements Model	Based es, Flow ing for	Modelin v-Oriente WebApp
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Requirement Modeling UNIT II Design (Convention Convention Convent	ements Modeling: Flow, Behavior, Patterns, WebApps – Requirements Modeling, Creating a Behavioral Model, Patterns for Requirements Modeling, Requirements Modeli	concepts, Class- odeling Strategic irements Model esign Process, I Genres, Arcl Iapping using D tegic Issues, Te tegies for Web oplications — S Path Testing, C	Based es, Flow ing for V Design nitectura ata Flow est Strat Apps, V oftware Control	Modelin v-Oriente WebApp  9 Concepts al Styles v.  9 regies for Validation Testing Structure
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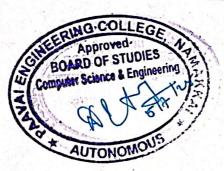
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#### **CO-POMAPPING:**

Mapping of Course Outcomes with Programme Outcomes: (1,2,3 indicates the strength of correlation) 3 – Strong, 2 – Mediu

			Programme Specific Outcomes (PSOs)										
PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2
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3	1	2.	1	1			•	1	2	1		1	1
3	1		1	2	-	. <del>.</del>	<i></i>	-	2	-		2	1
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CS233505	FOUNDATIONS OF DATA SCIENCE		0	
	OBJECTIVES	3 0		3
	the students to			
1 unc	derstand the data science fundamentals and its process.		41.7	alt g
2 des	scribe the data for building the model along and statistical basis for AI.	a de la companya de l		
3 ana	alyses the relationship between data using predictive model evaluation.			
4 util	lize the Python libraries for Data Wrangling.			4
5 pre	esent and interpret data using visualization libraries in Python.		9. J. H. L	i de
UNIT I	INTRODUCTION			
Data Science	ce: Benefits and uses - facets of data - Data Science Process: Overview -	Defining rese	arch go	als -
Retrieving building ap	data – Data preparation - Exploratory Data analysis – build the model oplications.	l- presenting f	indings	an
TINTE II	DESCRIPTION DATE.			
UNIT II	DESCRIBING DATA	Line Deservish	Table	
	Data - Types of Variables - Basic Statistical descriptions of Data - Descri			
	Describing Data with Averages - Describing Variability - Normal Distr	ributions and S	standar	d (2
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UNIT III	DESCRIBING RELATIONSHIPS  n –Scatter plots –correlation coefficient for quantitative data –computation	nal formula for	correl	7.7
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UNIT III Correlation coefficient				atio
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CO5	Apply visualization Libraries in Python to interpret and explore data	Analyzing (K4)
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# **CO-POMAPPING:**

Mapping of Course Outcome (CO's) with Programme Outcomes (PO's) and Programme Specific Outcomes PSO's

(1/2/3 indicates strength of correlation) 3-Strong, 2-Medium, 1-Weak

COs		4.35	Programme Specific Outcomes (PSOs)											
•	PO 1	PO 2	PO 3	PO4	PO 5	PO 6	PO 7	PO 8	PO 9	PO1 0	PO1 1	PO12	PSO1	PSO2
CO1	2	2	1	2	2	-,	( · · ).	- <del>-</del> 2.	1	1	1	2	2	2
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CO4	3	2	2	1	. 1	•	(Tear		1	_1	2	2	3	3
CO5	2	2	1	2	2	1.	1.5	NE :-	1	1	1	2	2	2



CS23506	DATA WAREHOUSING AND DATA MIN	VING LABORATORY	0 0	2	1
COURSE OB	JECTIVES				
To enable the	students to				
1. To in exerc	plement the concepts of data warehousing, OL ses.	AP, and data mining throu	igh pra	ctica	al
2. To pe	rform data preprocessing, transformation, and	analysis on real datasets.			
3. To ap	ply mining techniques such as classification, c	lustering, and association i	rule mi	ning	<u>.</u>
4	e data mining tools such as Weka, RapidMine based OLAP tools.	er, R, Python (Scikit-learn	, Pand	as),	or

# LIST OF EXPERIMENTS

- 1. Explore WEKA
- 2. Perform data preprocessing tasks
- 3. Demonstrate performing association rule mining on data sets
- 4. Implement the FP-Growth algorithm.
- 5. Demonstrate performing classification on data sets
- 6. Demonstrate performing clustering on data sets
- 7. Demonstrate performing Regression on data sets.
- 8. Credit Risk Assessment. Sample Programs using German Credit Data
- 9. Implement Apriori approach

	TOT	AL PERIODS	3 3	30
	SE OUTCOMES and of this course, the students will be able to		IAPPE est Leve	
CO1	Use data mining tools like WEKA, Python, or R to explore datasets perform basic operations.	and	ving (K	
CO2	Apply data preprocessing techniques such as cleaning, transformation and normalization to real-world datasets.	on, Apply	ing (K	3)
CO3	Implement and analyze association rule mining techniques including Apriori and FP-Growth algorithms.	g Analy	zing(K4	4)
CO4	Apply classification and clustering algorithms to discover patterns in datasets.	n Analy	zing(K4	4)

# CO PO MAPPING:

# Mapping of Course Outcome (CO's) with Programme Outcomes (PO's) and Programme Specific Outcomes (PSO's)

(1/2/3 indicates strength of correlation) 3-Strong, 2-Medium, 1-Weak

COs		PO's													
	. 1	2	3	4	5	6	7	8	9	10	11	12	1	2	
CO1	2	2	2	1	3	1 5 - Y	-		1	2	2	2	3	3	
CO2	2	3	2	-,	3	-	-		-	2		2	3	3	
CO3	3	3	2	2	3	-	-	11.		2	1	2	3	3	
CO4	3	2	3	2	3		0.00	-	·	2	2	2	3	3	

Approved
BOARD OF STUDIES
Computer Science & Engineering

CS235	507	ART	TIFICIAL INTELLIGENCE LABORATORY		0	0	4	2
COUI	RSE OBJ	<b>ECTIVES</b>						
To ena	able the st	udents to						
1 .	apply k	knowledge of va	rious methods of different problem solving and se	arching	technic	ques		
2			s of knowledge representation					
3	implem	ent unification a	and classification planning algorithm				<u> </u>	

# LIST OF EXPERIMENTS:

5

- 1. Implement basic search strategies 8-puzzle problem
- 2. Implement basic search strategies crypt arithmetic problem
- 3. Implement BFS and DFS for water jug problem

develop systems with logical reasoning

develop systems with probabilistic reasoning

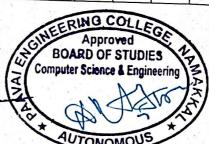
- 4. Implement a\* and memory bounded a\* search
- 5. Implement unification algorithm using python
- 6. Implement classical planning algorithm
- 7. Implement forward chaining and backward chaining using python
- 8. Implement Mini max algorithm for game playing (Alpha-Beta pruning)
- 9. Solve constraint satisfaction problems
- 10. Case Studies on NLP

COURSE OUTCOMES	TOTAL PERIODS 60
At the end of this course, students will be able to	BT Mapped
CO1:Design and implement search strategies	(Highest Level) APPLYING(K3)
CO2: Understanding the knowledge about the usage of predicate logic and first order logic for making inferences	APPLYING(K3)
CO3:Develop logic reasoning systems	APPLYING(K3)
CO4:Develop probabilistic reasoning systems	APPLYING(K3)

# CO's-PO's &PSO's MAPPING

CO's				Y. 1		PO's								PSO's		
	1	2	3	4	5	6	7	8	9	10	111	112	+-	F30	5	
1	3	3	3	2		-	<del>  •</del> •	-	-	10	11	12	1.1	2	3	
2	2	2		12		<u> </u>	-		-	1	2	1	3	2	1	
2	3	3	. 3	3					13.	3	2	3	3	3	. 2	
3	3	3	3	3	200		47 × 17	74. B.F	8	1	3	1	1	1	12	
4	3	3	3	2	1.532	4.5	19.41	7 - 7 -		2	+	1	1	1-	3	
5	3	3	3	2	0.250					- 2	3	1	2	2	1	
AVG	2	3		2					6.14	1	3	3	3	2	1	
	<u> </u>		3	3	Vo Com	en"			10,10	1	3	3	3	2	+-	

1-Low, 2-Medium, 3-High, '-' - No Correlation



UTONOMOUS -

GE23	501	PROFESSIONAL DEVELOPMENT III		0	0	2	1
COURS	E OBJ	DECTIVES		+	2 42		
To enabl	le stud	ents to					
•	enhanc	te their Resume writing skills and improving corporate vocabularies to	survive	in th	ie coi	porat	e world.
•	evalua	te their interview skills and improve their interview presentation.		5,			
•	solve t	he quantitative aptitude problems and improve their mental ability.					
•	improv	ve the critical thinking and reasoning skills.					
UNI	TI	RESUME WRITING SKILLS					6
Updated	Resum	e Building III – Self Introduction III – Dressing Etiquette – JAM V – Co	orporate	Voc	abula	ry	
UNIT	II	INTERVIEW SKILLS		ia i			6
Interview Interview UNIT		- General guidelines - Work Ethics - Group Discussion III - JAM VI - P  QUANTITATIVE APTITUDE	resentat	ion C	omp	etence	e – Mock
Cube Roo	ot and S	quare Root - Time and Work - Ages - Permutation and Combination -	Probabi	ility	- Cal	endar	
UNIT	IV	LOGICAL REASONING					9
Series Co	mpleti	on - Blood Relations - Coding and Decoding - Data Sufficiency - St	atement	ts an	d As	sumpt	ions
		TO	TAL PE	CRIC	DS:	1	30
		COMES on of the course, the students will be able to	BT M (High			)	
CO1	• exc	el in drafting Resumes and speaking.	C	Create	e (K	6)	
CO2	• dem	nonstrate the participative skills in group discussions and Interviews.	A	pplyi	ng (I	(3)	
CO3	• solv	re problems based on quantitative aptitude.	A	pplyi	ng (I	(3)	
CO4	• enh	ance their logical and verbal reasoning.	Aı	nalyz	zing (	K4)	

- 1. Aggarwal, R. S. A Modern Approach to Verbal & Non-Verbal Reasoning. Revised ed., 2024–25, S. Chand & Company Ltd., 2024.
- 2. Aggarwal, R. S. Objective General English: Fully Revised Video Edition. S. Chand & Company Ltd., 2022.

# REFERENCES

- 1. Abhijit Guha, "Quantitative Aptitude", Tata-Mcgraw Hill.2015.
- 2. Word Power Made Easy By Norman Lewis, Wr.Goyal Publications.2016.
- 3. Johnson, D.W. Reaching out Interpersonal Effectiveness and self- actualisation. Boston: Allyn and Bacon.2019.
- 4. Infosys Campus Connect Program students' guide for soft skills.2015.

CO/PC	MAPI	PING:												
		Mar (1/2	ping o 2/3 ind	f Cours	se Outo	come ((	CO's) w	vith Pro	gramr	ne Outc	omes (P	O's)		
CO's							4	Outcon			eat at			
003	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PS01	PS02
CO1	3	2	2	3	3	1	•	-	-				3	2
CO2	-	2	3	-	2		2	-	_		-	_	3	2
CO3	3	2	2	2	-	-	1	-	-				2	3
CO4	3	2	2	-	× <u>-</u>	1		-	-		2		2	3

