

## **MASTER OF COMPUTER APPLICATION**

Syllabus w.e.f. the Academic Session 2020-2021





MAULANA ABUL KALAM AZAD UNIVERSITY OF TECHNOLOGY WEST BENGAL

#### **Objective:**

To conduct software industry, corporate sector, academia, research-oriented MCA program following the AICTE model for MCA

#### **Eligibility:**

Candidates with the following eligibility can take admission in the 2-year MCA program approved by AICTE:

- A. Students who have passed Bachelor of Computer Application or Bachelor's degree in Computer Science Engineering or equivalent degree
- B. Students who have passed Bachelor of Science, Bachelor of Commerce or Bachelor of Arts with mathematics at 10+2 or at the graduation level with additional bridge courses as per the norms of the concerned university
- C. Candidates must have obtained at least 50 percent marks, or 45 percent marks in the case of candidates belonging to reserved categories, in the qualifying examination

#### **Duration:**

2 Years (4 Semesters)

#### **Program Educational Objectives (PEOs)**

- **PEO 01:** Technical Expertise: Develop the ability to plan, analyze, design, code, implement, test and maintain the software product for real time systems that are technically sound, economically feasible and socially acceptable
- **PEO 02:** Successful Career: Exhibit professionalism, ethical attitude with updated technologies in Computer Application based career and capability to set up their own enterprise in various sectors of Computer Applications
- **PEO 03:** Soft Skills: Develop communication skills, team work and leadership quality in their professional multidisciplinary projects and adapt to current trends by engaging in lifelong learning
- **PEO 04:** Life Long Learning: Prepare the students to pursue higher studies by acquiring knowledge in mathematical, computing and engineering principles in the field of computing and related fields and to work in the fields of teaching and research

#### Program Specific Outcomes (PSOs)

The post-graduates of Master of Computer Application Program will demonstrate:

- **PSO 01:** Software System Design and Development: The ability to apply software development life cycle principles to design and develop the application software that meets the automation needs of society and industry.
- **PSO 02:** Computing and Research ability: The ability to employ modern computer languages, environments and platforms in creating innovative career paths in SMAC (Social, Mobile, Analytics and Cloud) technologies.
- **PSO 03:** Professionalism and Ethics: Efficient team leaders, effective communicators and capable of working in multi-disciplinary environment following ethical values.

#### **Program Outcomes (POs)**

On Completion of MCA program, the post-graduates are expected to

- **PO 01:** Engineering Knowledge: Ability to apply knowledge of computing, science, mathematics and engineering fundamentals appropriate to the discipline
- **PO 02: Problem Analysis:** Ability to identify, critically analyze, formulate the computing requirements appropriate to its solution and develop computer applications
- **PO 03: Design/Development of Solutions:** Ability to design, implement and evaluate a computer-based complex system, process, component, or program to meet desired needs with appropriate consideration for public health and safety, cultural, societal and environmental considerations
- **PO 04:** Conduct Investigations of Complex Problems: Use of research-based knowledge and research methods including design of experiments, analysis and interpretation of data, and synthesis of the information to provide valid conclusions and develop Software with complete satisfaction to the Customer.
- **PO 05:** Modern Tool Usage: Ability to apply current technologies, skills, and modern IT tools necessary for computing practice with an understanding of the limitations.
- **PO 06:** The Engineer and Society: Ability to understand the impact of system solutions in a contemporary, global, economical, environmental and societal context for sustainable development.
- **PO 07:** Environment and Sustainability: Ability to understand the impact of the professional engineering solutions in societal and environmental contexts, and demonstrate the knowledge of, and need for sustainable development.
- **PO 08:** Ethics: Ability to discharge their duties with professional and ethical responsibilities as an individual as well as in multidisciplinary teams with positive attitude.
- **PO 09:** Individual and Team Work: Ability to function individually in effective manner and on teams, including diverse and multidisciplinary, to accomplish a common goal.
- **PO 10:** Communication: Ability to communicate effectively with a range of audiences and be customer friendly.
- **PO 11: Project Management and Finance:** Demonstrate knowledge and understanding of the engineering and management principles and apply these to one's own work, as a member and leader in a team to manage projects and in multidisciplinary environments and should be economically feasible.
- **PO 12:** Life-Long Learning: Recognize the need for, and have the preparation and ability to engage in independent and life-long learning in the broadest context of technological changes.

#### **Program Structure:**

	THEO	RY	PRAC	ΓICAL	SESSI	ONAL	Semester
SEMESTER	Courses	Credits	Courses	Credits	Courses	Credits	Credits
		[A]		[B]		[C]	[A+B+C]
Ι	4(C) + 1(E)	19	3	6	-	-	25
II	4(C) + 1(E)	19	3	6	-	-	25
III	3(C) + 2(E)	18	1	2	1	5	25
IV	1(0)	3	-	-	2	22	25
	ТС	TAL CR	EDIT→				100
* C →	Compulsory (	Courses					
* E →	Elective Cour	ses					
* 0→	Open Elective	e Courses					

#### **Project: Dissertation + Presentation + Project viva**

#### Session:

- Odd Semester/ 1<sup>st</sup> and 3<sup>rd</sup>: July December
- Even Semester/2<sup>nd</sup> and 4<sup>th</sup>): January June
- Lecture Hour: 1 Hour
- Subject wise Lecture per Week: 4

#### **Examination System:**

Subject wise Total Marks:	100
Semester Grade Point Average:	SGPA
Yearly Grade Point Average:	YGPA
Degree Grade Point Average:	DGPA

#### **Teaching Methodology:**

Lecture, Discussion, Presentation, Case Studies, Group Task, Assignment, Projects, Special Lecture by Industry Professionals

#### **General Guidelines:**

The 2-year MCA curriculum will be applicable w.e.f. the academic year 2020 - 2021. All rules and regulation regarding admission, examination, registration, migration and others shall exist according to MAKAUT norms.

### PART – I COURSE STRUCTURE

		Semester – I					
THF	EORY						
S1.	Paper Code	Paper Name	Con	tact H	ours	/Week	Credit
No	Faper Code	raper Maine	L	Т	P	Total	Clean
1	MCAN-101	Programming Concept with Python	3	1	-	4	4
2	MCAN-102	Relational Database Management System	3	1	-	4	4
3	MCAN-103	Computer Organization and Architecture	3	1	-	4	4
4	MCAN-104	Discrete Mathematics	3	1	-	4	4
5	Elective I		3	-	-	3	3
	MCAN-E105A	Environment and Ecology		1		1	1
	MCAN-E105B	Management Accounting					
	MCAN-E105C	Constitution of India					
	MCAN-E105D	Stress Management through Yoga					
	MCAN-E105E	Ethics in Business Profession					
	MCAN-E105F	Managerial Economics					
PRA	CTICAL						
1	MCAN-190	Soft Skill and Interpersonal Communication	-	-	4	4	2
2	MCAN-191	Python Programming Lab	-	-	4	4	2
3	MCAN-192	Relational Database Management System Lab	-	-	4	4	2
	Total Week	ly Contact Hours and Credit			•	31	25
BRI	DGE COURSE					1	1
[Onl	y for Students of	f Category "B" stated in the "Eligibilit	ty" Sec	tion]			
		Online Course on Fundamentals of 'C			ience'	or 'Co	mputer
Annl	lightion' on Infom	motion Tachnology' or so	-				-

Application' or 'Information Technology' or so

		Semester - II					
THE	ORY						
S1.	Paper Code	Paper Name	Con	tact H	ours	/ Week	Credi
No.		· · · · · · · · · · · · · · · · · · ·	L	Т	P	Total	
1	MCAN-201	Data Structure with Python	3	1	-	4	4
2	MCAN-202	Operating System	3	1	-	4	4
3	MCAN-203	Object Oriented Programming with JAVA	3	1	-	4	4
4	MCAN-204	Networking	3	1	-	4	4
5	Elective II		3	-	-	3	3
	MCAN-E205A	Numerical and Statistical Analysis					
	MCAN-E205B	Computer Graphics					
	MCAN-E205C	Probability and Statistics					
	MCAN-E205D	Introduction to Cyber Security					
	MCAN-E205E	Introduction to IoT					
	MCAN-E205F	Automata Theory and Computational Complexity					
PRA	CTICAL		•				
1	MCAN-291	Data Structure Lab with Python	-	-	4	4	2
2	MCAN-292	Operating System Lab (Unix)	-	-	4	4	2
3	MCAN-293	Object Oriented Programming Lab using JAVA	-	_	4	4	2
	<b>Total Weekly</b>	Contact Hours and Credit				31	25

#### [Only for Students of Category "B" stated in the "Eligibility" Section]

A minimum 8-week Online Course on Fundamentals of 'Software Engineering' or 'Systems Analysis and Design' or 'Business Systems Applications' or so

		Semester – III					
THE	EORY						
S1.	Paper Code	Paper Name	Con	tact H	ours /	Week	Credit
No.	Paper Code	raper Name	L	Т	P	Total	Credit
1	MCAN-301	Software Engineering using UML	3	1	-	4	4
2	MCAN-302	Artificial Intelligence	3	1	-	4	4
3	MCAN-303	Design and Analysis of Algorithm	3	1	-	4	4
4	Elective III		3	-	-	3	3
	MCAN-E304A	Image Processing		1		1	
	MCAN-E304B	Web Enabled JAVA Programming					
	MCAN-E304C	Cloud Computing	-				
	MCAN-E304D	Web Technology using PHP	-				
	MCAN-E304E	Android Application Development	-				
	MCAN-E304F	Basic Data Science	-				
5	Elective IV		3	-	-	3	3
	MCAN-E305A	Information Retrieval		1		1	
	MCAN-E305B	Data Warehousing and Data Mining					
	MCAN-E305C	Introduction to Big Data Analytics	-				
	MCAN-E305D	Graph Theory					
1	MCAN-E305E	Operation Research and	-				
		Optimization Techniques					
1	MCAN-E305F	Pattern Recognition	1				
	MCAN-E305G	Machine Learning	1				
PRA	CTICAL	1					
1	MCAN-E394	Elective III Lab	-	-	4	4	2
	(A/B/C/D/E/F)						
SES	SIONAL						
1	MCAN-381	Minor Project and Viva-voce	-	-	8	8	5
	Total Week	ly Contact Hours and Credit		1	1	30	25

		Semester IV					
THE	ORY						
Sl.			Cont	act H	ours /	'Week	C 1'
No.	Paper Code	Paper Name	L	Т	P	Total	Credit
1	<b>Open Elective</b>						
	MCAN-O401	Open Elective         • Business Analytics         • Robotics         • Bioinformatics         • Information Theory &Coding         • Automation in VLSI Design         • Intelligent Control         • Design of Embedded Systems         • Machine Learning         • Soft Computing         • Information Retrieval         • Distributed System         • Digital Marketing         *While opting for a domain for pursuing the Open Elective course, a student needs to ensure that the domain was not covered in previous semesters of the program.	-	-	-	-	3
SESS	SIONAL						
1	MCAN-481	Compressive Viva-voce	-	-	-	-	2
2	MCAN-482	Major Project and Viva-voce	-	-	28	28	20
	Total Week	y Contact Hours and Credit			1	28	25



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### **First Year: Semester-I**

Code M(	NANI 101	Danau Duaguamming	Concept with Duthon	
Code: MC	AN-101 Hours / Week: 4	Paper: Programming Total Contact Hours:		Credit: 4
Contacts Course O		Total Contact Hours:	40	Creuit: 4
		f this course, students will b	e able to:	
		d comprehend the concept of		
		olve simple programming p		
		mber syntax and semantics	of Python.	
		ing secondary storage.		
		library for data analysis.	4	
		ment different solutions for	the same problem and analyze why on	e solution is better than the
oth				
	o write program for	real life problem.	COUDCE CONTENT	
UNIT	E		COURSE CONTENT	
	Fundamentals of			(6L)
			mputer System, Primary & Secondary	
			ly language, High level language, Con	
1			adecimal) with signed and unsigned	d numbers (using 1's and 2's
			n and arithmetic operations.	
		ked BCD system, ASCII. I	EEE-754 floating point representation	(half- 16 bit, full- 32 bit, double-
	64 bit).			
	Programming Ba			(2L)
2			Pseudo codes, structured programming	ng, Example of Flowchart and
	Algorithm represe			
	Variable and Ex			(4L)
3			(arithmetic and logical) and their eva	
5	precedence). Assi	gnment operation; differen	ce between left hand side and right ha	and side of assignment, Console
	input/output: takin	ng input from user and print	ing user information.	
	Control Stateme	nt and Iteration		(5L)
4			ments within if, multiple if statement.	
		g Loops using Break and C	ontinue, Else Statement, Range Stateme	ent and Pass Statement in Loop.
5	Collections			(2L)
	Strings, List, Tup	les, Dictionary, Set, Selection	on sort, Bubble sort	
-	Function			(2L)
6			tion passing values, function returning	values, default parameter values,
	Recursive functio			( <b>4T</b> )
7	File Managemen			(4L)
7			putes, encoding, closing), read() & v	write() methods, tell() & seek()
		g & deleting files and direc	tories	
8	Errors and Exce		ling and the second for the second f	(2L)
			ling exceptions with try/except, Cleani	
	Classes and Obj			(5L)
0			nction, Methods, Self Parameter, Mod	
9			nent, Inheritance and Polymorphism, and Date Objects, RegEx Module a	
	Handling.	violule, would datetime	and Date Objects, RegEx Module a	and Regex Functions, Exception
	Modules& Packa	Agos		(2L)
10		le, Creating module, Funct	ion aliases nackages	(2L)
	Numpy	ie, creating module, i une	ion anases, packages	(6L)
11		reading files exploratory	lata analysis, data preparation and pro	
11		t, Histogram, Box plot, Pair		cessing, , matpiono. Seatterpiot,
Reference		t, motogram, Box piet, i an	piot	
		of Computer Fundamenta	s, Khanna Publishing House	
		ng Python by Programming		
		The Complete Reference Py		
		nscroft, S. Holden, Python in		
•	Jason Rees-Python	Programming:Practical intro	oduction to Python Programming for to	
			By Step Guide From Beginner To Expe	ert (Beginner,
	Intermediate & Adv	anced)		W.C.
•	Mark Pilgrim-Diva	Into Python, Springer-Verla	g Berlin and Heidelberg GmbH & Co.	KG
•	Summerfield Mark-	Programming in Python 3,	Pearson Education India	

Code: M	CAN-102	Paper: Relational Database Management System
	Hours / Week: 4	Total Contact Hours: 40 Credit: 4
Course C		
		this course, students will be able to:
		database over the file system.
		ment the process of data insertion, retrieval, and manipulation.
		ze the functional dependencies among attributes of the entity set and normalization between
	the relations.	
		pt for a database transaction.
	1 ~	ment the Transaction control and concurrency control management.
		tables, PL/SQL programs, triggers, database files, indexing of RDBMS.
UNITS		COURSE CONTENT
011115	Basic Concept	(7L)
		nent System, File based system, Advantages of DBMS over file based system, Database
		1 DBMS Architecture, Three level architecture of DBMS or logical DBMS architecture,
		el architecture, Physical DBMS Architecture, Database Administrator (DBA) Functions &
		•
1		ices and Data Dictionary Relational and ER Modela, Data Modela, Relational Model, Damaing, Turala and Relation
1		Relational and ER Models: Data Models, Relational Model, Domains, Tuple and Relation, late keys, Primary keys and foreign key for the Relations, Relational Constraints, Domain
	1 2	
		nstraint, Integrity Constraint,- Update Operations and Dealing with Constraint Violations,
	Relational Operation	
		<b>p</b> (ER) Model: Entities, Attributes, Relationships, More about Entities and Relationships,
		Diagram to Relational Database.
		/ And Normalization (8L)
		e Integrity, The Keys, Referential Integrity, Entity Integrity, Redundancy and Associated
2		alued Dependencies, Normalization, Rules of Data Normalization, The First Normal Form,
		al Form, The Third Normal Form, Boyce CODD Normal Form, The Fourth Normal Form,
		al Form, Multi-valued Functional Dependency, Attribute Preservation, Losslessjoin
		pendency Preservation.
	File Organization	(4L) Design Issues, Storage of Database on Hard Disks, File Organization and Its Types, Heap
3		les), Sequential File Organization, Indexed (Indexed Sequential) File Organization, Hashed
5		Types of Indexes, Index and Tree Structure, Multi-key File Organization, Need for Multiple
		i-list File Organization, Inverted File Organization.
	Structured Query	
4		nmands, Data Definition Language, Data Manipulation Language, Data Control Language,
		ol Language, Queries using Order by, Where, Group by, Nested Queries. Joins, Views,
		and Synonyms, Table Handling.
		Concurrency Management (8L)
5		rurrent Transactions, Locking Protocol, Serializable Schedules, Locks Two Phase Locking nd its Prevention, Optimistic& Pessimistic Concurrency Control. Database Recovery and
5		Recovery meaning, Kinds of failures, Failure controlling methods, Database errors, Backup
		ques, Security & Integrity, Database Security Authorization.
(	PL/SQL	(6L)
6		SQL, Variables & Data types, Basic blocks, Conditional & branching statement, Handling
D C		Function, Procedure, Package and Exception.
Reference		
•		Sudarshan-Data Base System Concepts, MH.
•		indamentals of Database Systems, Pearson ition to Database, Addison-Wesley Publishing Company
•		haryya-Data Base Management Systems, TMH
•		/SQL Programming,SPD/O'REILLY
•		agement Systems, VIKAS
•		Processing:Fundamentals, Design &Implementation,PHI
•		PL/SQL for Oracle 8 & 8i, Wiley Dreamtech
•		G. Singh- Concepts of Database Management System, Kalyani Publishers

Master of C	Computer Application
Code: MC	CAN-103 Paper: Computer Organization and Architecture
	Hours / Week: 4 Total Contact Hours: 40 Credit: 4
Course Ou	
	essful completion of this course, students will be able to:
	scribe the merits and pitfalls in computer performance measurements and analyze the impact of instruction set
	hitecture on cost-performance of computer design
	plain Digital Logic Circuits, Data Representation, Register and Processor level Design and Instruction Set
arc	hitecture
🖌 🖌 Sol	ve problems related to computer arithmetic and Determine which hardware blocks and control lines are used for
spe	cific instructions
✓ De	sign a pipeline for consistent execution of instructions with minimum hazards
✓ Exp	plain memory organization, I/O organization and its impact on computer cost/performance.
UNITS	COURSE CONTENT
	INTRODUCTION (8L)
	Digital Logic Design: Axioms and laws of Boolean algebra, Reduction of Boolean expressions, conversion
1	between canonical forms, Karnaugh map (4 variable), Half Adder, full adder, 4-bitparallel parity bit generator,
	checker circuit, Decoder, Encoder, Multiplexer, IC RAM, ROM, Memory Organization, Sequential Circuits, State
	transistors, Flip-flop, RS, JK, D-Latch, Master-slave.
	INSTRUCTION SET ARCHITECTURE: (8L)
	Memory Locations and Addresses: Byte Addressability, Big-Endian and Little-Endian
2	Assignments, Word Alignment, Instructions and Instruction Sequencing, Addressing Modes,
	Assembly Language, Subroutines, Additional Instructions, dealing with 32-Bit Immediate
	Values.
	BASIC PROCESSING UNIT & PIPELINING (8L)
	Basic Processing Unit: Some Fundamental Concepts, Instruction Execution, Hardware Components, Instruction
3	Fetch and Execution Steps, Control Signals, Hardwired Control, CISC-Style Processors.
	Pipelining: Basic Concept, Pipeline Organization, Pipelining Issues, Data Dependencies, Memory Delays,
	Branch Delays, Pipeline Performance Evaluation.
	MEMORY ORGANIZATION (8L)
4	Basic Concepts, Semiconductor RAM Memories, Read-only Memories, Direct Memory Access, Memory
	Hierarchy, Cache Memories, Performance Considerations, Virtual Memory, Memory Management Requirements,
	Secondary Storage. INPUT OUTPUT & PARALLEL PROCESSING (8L)
	Basic Input Output: Accessing I/O Devices, Interrupts, Input Output Organization: Bus Structure, Bus
5	Operation, Arbitration, Interface, Interconnection Standards. Parallel Processing: Hardware Multithreading,
5	Vector (SIMD) Processing, Shared-Memory Multiprocessors, Cache Coherence, Message-Passing
	Multicomputers, Parallel Programming for Multiprocessors, Performance Modeling.
Reference H	
	Computer Organization and Embedded Systems, 6 <sup>th</sup> Edition, Hamacher Carl, et. al, Tata McGraw
]	Hill, New Delhi, 2011.
• (	Computer Organization and Design: The Hardware Software / Interface, 5th Edition, 1994, Patterson David A.
• (	Computer System Architecture, Revised 3 <sup>rd</sup> Edition, Mano M. Morris, Pearson Education,

	AN-104 Paper: Discrete Mathematics	
Contacts H	Iours / Week: 4     Total Contact Hours: 40	Credit: 4
	ssful completion of this course, students will be able to:	
	erpret the problems that can be formulated in terms of graphs and trees.	
	plain network phenomena by using the concepts of connectivity, independent sets, cliques,	matching, graph coloring
✓ Ac	hieve the ability to think and reason abstract mathematical definitions and ideas relance ncepts of well-ordering principle, division algorithm, greatest common divisors and congrue	
🗸 Ap	oply counting techniques and the crucial concept of recurrence to comprehend the control of the	
	alyze the logical fundamentals of basic computational concepts.	
	mpare the notions of converse, contrapositive, inverse etc. in order to consolidate the com	prehension of the logica
	otleties involved in computational mathematics.	prenention of the rogics
UNITS	COURSE CONTENT	
1	Logic and Proofs	(3L)
1	Propositional logic, Propositional equivalences, Predicates and quantifiers, Nested quantif	
2	<b>Principles of Mathematical Induction</b> The Well-Ordering Principle, Recursive definition, The Division algorithm: Prime Number Divisor: Euclidean Algorithm, The Fundamental Theorem of Arithmetic.	(5L) ers, The Greatest Comm
3	<b>Sets and Sequence</b> Sets, Relation and Function: Operations and Laws of Sets, Cartesian Products, Binary Relation, Equivalence Relation, Image of a Set, Sum and Product of Functions, Biject Composite Function, Size of a Set, Finite and infinite Sets, Countable and uncountable argument and The Power Set theorem, Schroeder-Bernstein theorem. Fuzzy set, Basic pro-	ive functions, Inverse a le Sets, Cantor's diagor
4	<b>Counting and Combinatorics</b> Counting, Sum and product rule, Principle of Inclusion Exclusion. Pigeon Hole Principle Double Counting. Linear Recurrence relations - methods of solutions. Generating Fun Combination.	<b>(8L</b> e, Counting by Bijection
5	Algebraic Structure Algebraic Structures with one Binary Operation, Semi Groups, Monoids, Groups, Quotient Structures, Free and Cyclic Monoids and Groups, Permutation Groups, Substruct Algebraic Structures with two Binary Operation, Rings, Integral Domain and Fields. Boo Ring, Identities of Boolean Algebra, Duality, Representation of Boolean Function, Dis Normal Form	ctures, Normal Subgroup blean Algebra and Boole
6	<b>Graph and Tree</b> Graphs and their properties, Degree, Connectivity, Path, Cycle, Sub Graph, Isomorphism, Walks, Graph Colouring, Colouring maps and Planar Graphs, Colouring Vertices, Colour Perfect Graph, definition properties and Example, rooted trees, trees and sorting, weight Bi-connected component and Articulation Points, Shortest distances.	ing Edges, List Colourin

- DeoNarsingh Graph Theory With Applications To Engineering And Computer Science, PHI Learning Arumugam, Ramachandran- Invitation to Graph Theory, Scitech Publications (India) •
- •

Master of C	Computer Application
	AN-E105A Paper: Environment and Ecology
	Hours / Week: 3Total Contact Hours: 30Credit: 3
Course Ou	
	essful completion of this course, students will be able to:
	e able to understand the natural environment and its relationships with human activities.
	e able to apply the fundamental knowledge of science and engineering to assess environmental and health risk. e able to understand environmental laws and regulations to develop guidelines and procedures for health and
	fety issues
	e able to solve scientific problem-solving to air, water, noise and land pollutions.
UNITS	COURSE CONTENT
	Introduction         (4L)           Basic ideas of environment and interrelationship among man society and environment.         (4L)
1	Environmental problems and issues, Segments of environments, Natural Cycles of environments
	Mathematics of population growth and its associated problems, Logistic population growth
•	Elements of Ecology (3L)
2	Open and closed system ecology, species, population, community, definition of ecosystem-components types
	and functions, Environmental perspectives, Montreal protocol
2	Pollutants and Contaminants (3L)
3	Definition of primary and secondary pollutants and contaminants. Source and effects of different air pollutants
	suspended particulate matter, oxides of carbon, nitrogen, sulphur particulate (5L)
4	Structures of the atmosphere, global temperature models, Greenhouse effect, global warming; acid rain: causes, effects and control. Lapse rate and atmospheric stability; pollutants and contaminants; smog; depletion of ozone
	layer; standards and control measures of air pollution.
	Water Pollution       (5L)
	Hydrosphere; pollutants of water: origin and effects; oxygen demanding waste; thermal pollution; pesticides;
5	salts. Biochemical effects of heavy metals; eutrophication: source, effect and control. Water quality parameters:
	DO, BOD, COD. Water treatment: surface water and wastewater.
6	Land Pollution (5L)
0	Land pollution: sources and control; solid waste: classification, recovery, recycling, treatment and disposal.
	Noise Pollution (5L)
7	Noise: definition and classification; noise frequency, noise pressure, noise intensity, loudness of noise, noise
	threshold limit value; noise pollution effects and control.
Reference H	
• ]	Basic Environmental Engineering and Elementary Biology, GourKrishna Das Mahapatra, Vikas
	Publishing House P. Ltd. Environmental Chemistry, A. K. De, New Age International.
• ]	Environmental Engineering, G.M.Masters, Tata Mc Graw Hills
	Environmental Chemistry with Green Chemistry, A. K. Das, Books and Allied P. Ltd.
	Fundamentals of Environment & Ecology, D. De, D. De, S. Chand & Company Ltd.

Contacts Hours / Week: 3       Total Contact Hours: 30         Course Outcome:       After successful completion of this course, students will be able to:         ✓ Understand the basic concepts related to Business.       ✓ Demonstrate the roles, skills and functions of different discipline of business management	
✓ Understand the basic concepts related to Business.	
· Demonstrate the roles skills and functions of different dissipling of husiness managemen	
	banking system
$\checkmark$ To disseminate knowledge among the students inculcate with theoretical structures about	
$\checkmark$ Record basic accounting transactions and prepare annual financial statements; and analys	e, interpret and communicate
the information contained in basic financial statements	
✓ Analyse and provide recommendations to improve the operations of Organisations throug Management accounting techniques	sh the application of Cost and
✓ Equip students with in-depth and expert knowledge of Tally ERP with GST.	
UNITS COURSE CONTENT	
Introduction	(3L)
<sup>1</sup> Basics of management; Planning, scheduling, organizing, staffing, directing, contro	
Management	(3L)
2 Marketing Management, Financial management, Operation management,	· · · · · · · · · · · · · · · · · · ·
Human resource management, Management information System	
Strategy	(3L)
3 Firm and its environment, strategies and resources, industry structure and analy	sis, corporate strategies and it
evaluation, strategies for growth and diversification, strategic planning	
<b>Business Trade and Banking</b> Business: Types of business, Sole Proprietorship, Partnership, Limited company	(3L)
characteristics.	and cooperative society – the
4 Banking: role of commercial banks; credit creation and its importance in indust	rial functioning Role of centra
banking. For or commercial canno, creat creation and its importance in maast bank: Reserve Bank of India.	in interesting. Role of contra
International Business or Trade Environment.	
Financial Accounting	(7L)
5 Journals, Ledgers, Trial Balance, Profit & Count, Balance Sheet, Finan	
Financial Statement Analysis and Interpretation (Financial Ratio and Cash Flow and	• /
Cost Accounting	(7L)
6 Concepts and Classification of costs, Cost Sheet	
Break Even Analysis, Variance Analysis, Cost-volume profit (CVP) relationship, C	
7 Packages Financial accounting computer package (Tally ERP with GST)	(4L)
Reference Books:	
• Financial Accounting- A Managerial Perspective, R. Narayanswami, Prentice-Hall	of India Private Limited New
Delhi	
• Fundamentals of Financial Management, Horne, James C Van, Prentice-Hall of India	Private Limited, New Delhi
Modern Economic Theory, H. L. Ahuja., S. Chand. New Delhi.	
Management Accounting, Khan & Jain, TMH	

Management Accounting, M.E. ThukaramRao, New Age International

	CAN-E105C Paper: Constitution of India Hours / Week: 3 Total Contact Hours: 30	Credit: 3
Course O		
	cessful completion of this course, students will be able to:	
	Understand the premises informing the twin themes of liberty and freedom from a civ	vil rights perspective.
	To address the growth of Indian opinion regarding modern Indian intellectuals' cor	
	civil and economic rights as well as the emergence of nationhood in the early years o	
	To address the role of socialism in India after the commencement of the Bolshevik	
	on the initial drafting of the Indian Constitution.	1
UNITS	COURSE CONTENT	
	History of Making of the Indian Constitution	(5L)
1	History Drafting Committee, (Composition & Working)	
	Philosophy of the Indian Constitution	(5L)
2	Preamble Salient Features	
	Contours of Constitutional Rights & Duties	(5L)
	Fundamental Rights, Right to Equality, Right to Freedom, Right against Ex	
3	Religion, Cultural and Educational Rights, Right to Constitutional Remedies, Di	
	Fundamental Duties.	rective Trinciples of State Toney
	Organs of Governance	(5L
	Parliament, Composition, Qualifications and Disqualifications, Powers and	
4	Governor, Council of Ministers, Judiciary, Appointment and Transfer of Jud	
	Functions	
	Local Administration	(5L)
	District's Administration head: Role and Importance, Municipalities: Introduc	
5	Representative, CEO of Municipal Corporation. Pachayati raj: Introduction, PR	
	and their roles, CEO ZilaPachayat: Position and role. Block level: Orga	inizational Hierarchy (Differen
	departments), Village level: Role of Elected and Appointed officials, Importance of grass root democracy	
	Election Commission	(5L
6	Role and Functioning. Chief Election Commissioner and Election Commissioners. State Election Commission:	
	Role and Functioning. Institute and Bodies for the welfare of SC/ST/OBC and we	omen.
eference	Books:	
	The Constitution of India, 1950 (Bare Act), Government Publication.	
	Dr. S. N. Busi, Dr. B. R. Ambedkar framing of Indian Constitution, 1st Edition, 201	.5.
	M. P. Jain, Indian Constitution Law, 7th Edn., Lexis Nexis, 2014.	
•	D.D. Basu Introduction to the Constitution of India Lexis Nexis 2015	

• D.D. Basu, Introduction to the Constitution of India, Lexis Nexis, 2015.

Code: MO	CAN-E105D	Paper: Stress Management through Yoga		
Contacts Hours / Week: 3		Total Contact Hours: 30	Credit: 3	
Course O	outcome:			
After succ	cessful completion of this	course, students will be able to:		
✓ T	To achieve overall health	of body and mind		
√ T	To overcome stress			
UNITS		COURSE CONTENT		
1	Astanga		(8L)	
1	Definitions of Eight parts of Yoga (Ashtanga)			
	Yam and Niyam		(8L)	
2	Do's and Don't's in	life. i) Ahinsa, satya, astheya, bramhacharya and aparigraha ii)	Shaucha, santosh, tapa	
	swadhyay, ishwarprani	Idhan	-	
	Asan and Pranayam		(8L)	
3		and their benefits for mind & body ii)Regularization of breathing te		
	Typesof pranayama		1	
4	Meditation Techniqu	es	(6L)	
Reference	Books:		`````````````````````````````````	
•	Janardan Swami Yogabh	yasi Mandal- Yogic Asanas for Group Tarining-Part-I, Nagpur		
•	Swami Vivekananda- Ra	ajayoga or conquering the Internal Nature, AdvaitaAshrama (Public	cation Department),	
	Kolkata		- //	

	CAN-E105E	Paper: Ethics in Business Profession	
	Hours / Week: 3	Total Contact Hours: 30	Credit: 3
Course O			
		urse, students will be able to:	
		work ethics, Learn to respect others and develop civic virtue.	
		onsibilities of the engineers, create awareness about the customs a	and religions, Install
		Loyalty and to appreciate the rights of others.	
		ecome a social experimenter, Provide depth knowledge on framing	of the problem and
	etermining the facts.		
		v, risk & risk benefit analysis, Provide knowledge on Intellectual Pro	
		bal issues, Create awareness on computer and environmental ethics,	Analyze ethical
	roblems in research.		
UNITS		COURSE CONTENT	
	Human Values		(6L)
1		ics-Integrity-Work Ethic-Service learning, Civic Virtue, Respect	
		g, Honesty, Courage-Cooperation, Commitment, Empathy, Self Con	nfidence Character.
	<b>Professional Ethics</b>		(6L)
2		Ethics-Variety of moral issued, Types of inquiry, Moral dilemmas	
-		n's theory, Consensus and controversy, Models of professional re-	oles, Theories about
	right action, Self-interest,		
	Professional As Social E		(6L)
3		erimentation, Framing the problem, Determining the facts, Codes	
		ues, Common Ground, General Principles, Utilitarian thinking respec	
	Safety, Responsibilities		(6L)
4		sment of Safety and Risk – Risk Benefit Analysis and Reducing	
		argaining – Confidentiality – Conflicts of Interest – Occupational C	rime – Professional
	Global Issues	– Intellectual Property Rights (IPR) – Discrimination	((1)
		ure issues-Environmental Ethics, Computer Ethics -Computers a	(6L)
5		puters as the object of Unethical acts, Autonomous Computers,	
5		, Code of Conduct, Corporate Social Responsibility. Ethics and I	
	Ethical Problems in resear		cesearen, maryzing
Reference			
		S, Senthil Kumar V. S, "Engineering Ethics", Prentice Hall of India,	New Delhi.
•	A. R. Aryasri, Dharaniko	aSuyodhana "Professional Ethics and Morals" Maruthi Public	ations.
•	Mike W. Martin and Roland	Schinzinger, "Ethics in Engineering", Tata McGraw Hill, New Dell	ni.
•	John R Boatright, "Ethics	and the Conduct of Business", Pearson Education, New Delhi	

Code: M	ICAN-E105F	Paper: Managerial Economics	
	s Hours / Week: 3	Total Contact Hours: 30	Credit: 3
	Outcome:	Total Contact Hours, 50	cituit b
		course, students will be able to:	
		s of managerial economics.	
	To understand and interpr		
		s between short-run and long-run costs.	
		betitive markets including substitution.	
		g and how it relates to price discrimination and total revenue.	
		consequences of different market conditions.	
		f price and output decisions of firms under various market struct	ture.
UNITS	<u> </u>	COURSE CONTENT	
	Introduction (2L)		
1		gerial Economics, Basic problems of an economic system;	Goals of managerial decision
1	making; Resource allo		Could of managemat accision
	Demand Analysis (61	•	
		- Law of Demand, Explaining the law of demand, Violations o	of theLaw of Demand. Shifts ir
		Demand: Price Elasticity (at a point andover and interval), Fa	
		hange in TotalRevenue, AR, MR and Price elasticity, Range	
2		rior, Superior and Normal goods, Income Elasticity and Shar	
		itutes and Complements	1
		udget line and consumer equilibrium	
		ls of demand estimation (concepts only)	
	Production and Cost	· · · · · · · · · · · · · · · · · · ·	
		Short Run and Long Run, Production with One Variable Inp	ut, Total Product, Average and
		w of Variable proportions, Relationship between TP, AP and MI	
		Production, Fixed and Variable Costs, Short Run Total, Av	
		them, Short Run Cost Curves, Relationshipbetween AVC, MC	C, AP and MP; Long run cost
3	curves, Relationship b	etween LAC and SAC, Economies of Scale and Scope.	
		Variable Inputs, Isoquants – Characteristics, Marginal Rate of	
		, Laws of Returns to Scale, Isocost Curves, * # Finding the Opti	
		, Production of a given output at Minimum Cost, Production of	
		n a given level of Cost, Expansion Path, Finding the Long Run	
		he Production Function,	
4	Alternate Goals of M	anagerial Firms (2 L) Revenue maximization; Managerial utility maximization	
		Making under Alternative Market Structures (6 L)	
		rfect Competition, #Profit Maximization in Competitive Market	s Output Decision in the Short
		t, Short Run Supply for the Firm and Industry; Output Decision	
5		ly for the Perfectly Competitive Industry.	I in the Long Run, Dreak Liven
c	rom, zongrom sopp		
	Price and output decis	sion under different market structure – Monopoly, Monopolistic	
		y – cartel, price leadership.	
	Pricing Decisions [4]		
(		nder Monopoly, Transfer Pricing.	
6	Market Failure		
	Game theory &Asymr	netric information	
Reference			
•	Damodaran, Suma – Ma	nagerial Economics – Oxford University Press	
•	Lipsey & Chrystal – Ec	onomics – Oxford University Press	
•		nagerial Economics – Pearson Education.	
٠		- Micro Economics – Pearson Education	
•	H.L. Ahuza- Manageria		
•	D N Dwivedi- Manage	rial Economics. Prentice Hall.	

Code: MC	1 1	Cuality 2	
Contacts H Course Ou	Total Contact Hours: 40	Credit: 2	
	ssful completion of this course, students will be able to:		
	fectively communicate through verbal/oral communication and improve the listening skills		
	le to be self-confident with positive vibes		
	tively participate in group discussion / meetings / interviews and prepare & deliver presentations		
	come more effective individual through goal/target setting, self-motivation and practicing creative th	inking.	
	nction effectively in multi-disciplinary and heterogeneous teams through the knowledge of tea		
	sonal relationships, conflict management and leadership quality.	,	
UNITS	COURSE CONTENT		
	Soft Skills& Interpersonal Communication		
	An Introduction - Definition and Significance of Soft Skills; Process, Importance and Measuren	nent of Soft Ski	
1	Development.		
1	Inter personal relations; communication models, process and barriers; team communication; developing		
	interpersonal relationships through effective communication; listening skills; essential formal writing skills		
	corporate communication styles -assertion, persuasion, negotiation.		
	SWOT & Creative Thinking		
2	Discovering the Self; Setting Goals; Beliefs, Values, Attitude, Virtue.		
	Developing Positive Thinking and Attitude; Driving out Negativity; Meaning and Theories of Motivation		
	Enhancing Motivation Levels.		
	Corporate Communication		
	<b>Public Speaking:</b> Skills, Methods, Strategies and Essential tips for effective public speaking. <b>Group Discussion:</b> Importance, Planning, Elements, Skills assessed; Effectively disagree	aina Initiatin	
	Summarizing and Attaining the Objective.	eing, initiatin	
3	<b>Interview&amp; Presentation Skills:</b> Interviewer and Interviewee– in-depth perspectives. Before, Dur	ing and After th	
	Interview.		
	Tips for Success: Types, Content, Audience Analysis, Essential Tips – Before, During and At	fter, Overcomin	
	Nervousness.		
	Non-Verbal Communication & Personality Development		
4	Importance and Elements; Body Language.		
·	Concept, Essentials, Tipsc		
	Meaning, Nature, Features, Stages, Models; Learning Skills; Adaptability Skills.		
5	Business Etiquette & Team Work		
5	Concept of Teams; Building effective teams; Concept of Leadership and honing Leadership skills. Meaning, Nature, Features, Stages, Models; Learning Skills; Adaptability Skills.		
eference B			
	Janaging Soft Skills for Personality Development – edited by B.N.Ghosh,McGraw Hill India, 2012.		
	ffective Communication and Soft Skills, Nitin Bhatnagar, Pearson Education India, 2011		
	nglish and Soft Skills – S.P.Dhanavel, Orient Blackswan India, 2010.		

	of Computer Application			
Code: MC	AN-191 Iours / Week: 4	Paper: Python Pro Total Contact Hour		Credit: 2
Course Ou		Total Contact Hour	5. 10	
		ourse, students will be able t	to:	
		lating to different logical pr		
			ors reported by the compiler.	
		he native data types (Pythor		
	o implement conditional bi		,	
	o decompose a problem int			
		om and write into simple tex	t files.	
✓ To	o understand the basic con-	cept of OOPs		
✓ To	o understand and implement	nt Python NumpyArrray ope	erations	
UNITS		COUR	RSE CONTENT	
1	<b>Python Basics:</b> Installin Program	ng Python, Setting up Path	h and Environment Variables, 1	Running Python, First Python
2		ignment, Understanding D	, Identifiers, Python Statement Data Type, Data Type Conversi	
3	<b>Operators and Expres</b> Associative Operators.	sions: Operators in Python	n, Expressions, Precedence, As	ssociativity of Operators, Non
4		cision making statements, P	ython loops, Python control state	ements.
5	Python Native Data	<b>Types:</b> Numbers, Lists, Tu ir methods and operations).	uples, Sets, Dictionary, Functio	ons & Methods of Dictionary,
6			d functions, Anonymous funct	ions, Pass by value, Pass by
7	Exception Handling: E	ceptions, Built-in exception	ns, Exception handling, User def	ined exceptions in Python.
8	File Management in Py	thon: Operations on files (	opening, modes, attributes, enco ng files in Python, directories in	ding, closing), read() & write()
9		Object Class, Constructors,	Inheritance	
10	<b>Python Numpy</b> Numpy data types, Oper filter)	ations on Numpy Array (in	dexing,slicing, shape/reshape, it	eration, join, split, search, sort,

Code: MC	AN-192	Paper: Relational Database Management System Lab	
	Hours / Week: 4		edit: 2
Course Ou	itcome:		
After succe	essful completion of this	s course, students will be able to:	
		onship Diagram (ERD) model as a blueprint to develop the corresponding relational	model in
	RDBMS system like Or		
		f Structured query language (SQL) to create a relational database from scratch throug	gh
		s constraints in Oracle RDBMS system.	
1		of Structured query language (SQL) for storing and modification of data in Oracle RI	JBMS
	stem.		1 6 1 4
		f Structured query language (SQL) to construct complex queries for efficient retrieva	al of data
		per the user requirement specifications.	
		various P/L SQL concepts like cursor, trigger in creating database programs. atabase backend system using SQL and P/L SQL programming to establish overall in	nto onite.
	the database system.	atabase backend system using SQL and F/L SQL programming to establish overall in	megniy
		ion, Procedure and Package and Apply Exception.	
UNITS		COURSE CONTENT	
01115	Creation of a databa	ase based on given ERD Model:	
	SQL Data Definition		
		ble structure, Apply (and Alter) constraints on columns/tables viz., primary key, fo	oreign kev
		ck. Verify/ Review the table structure (along with applied constraints) using appro	
1		user_constraints, user_cons_columns, etc. Create view, materialized view using or	
	table.		
	SQL Data Manipulati	ion Language (DML)	
	Insert into rows (once	e at a time/ and in bulk) from a table, Update existing rows of a table, Delete rows (a	a few or al
	rows) from a table.		
	Data Query Langua		
		ere structure - Usage of Top, Distinct, Null keywords in query, Using String and	
		ng Where Clause with various Operators and logical combination of various	conditions
		der By clause. Usage of IN, LIKE, ALL keywords.	
		Natural Joins, equi-join, non-equi-join, Self-Join, Inner Join, Outer (left, right) Join.	
2	Set operations:		
2	Using single row fund	nus set operations on table data using SQL.	
		Idle ambiguity of null data), upper, lower, to_date, to_char functions, etc.	
		e row functions in Queries like Count, Sum, Min, Max, Avg, etc, using Group By a	and Having
		By with Rollup and Cube.	ind maving
		with various nested structure of Sub Queries - use in from or where clause with more	re than one
	1 7 0	elated sub-query- Ranking table data using correlated sub-query.	
	PL/SQL		
		d Functions- Basic programming constructs of PL / SQL like if, else, else-if, loop,	, while, fo
	structure		
	Populate stored proce	dure variables with the data fetched from table using SQL command.	
3		s - Creating Cursors, parameterized cursor, Locks on cursors, Exploring advantages	
5		rs - Constraints Vs Triggers, Creating, Altering, Dropping triggers, use of for/ after/	
		er to validate/ rollback a Transaction, Automatically populate integer data based pr	rimary key
	columns (e.g., Id.) usi		
	0	rocedure & Package – Create Function, Create Procedure and Create Package.	
	Exception Handling.		



## **MASTER OF COMPUTER APPLICATION**

Syllabus w.e.f. the Academic Session 2020-2021





MAULANA ABUL KALAM AZAD UNIVERSITY OF TECHNOLOGY WEST BENGAL

### **First Year: Semester-II**

Code: MC	1 0	~ ~ .
	Hours / Week: 4 Total Contact Hours 40	Credit: 4
Course O		
	essful completion of this course, students will be able to:	1.
	Understand the concept of abstract data type such as stack, queue, linked list, ar	id trees
	Those appropriate data structure to design algorithm to solve the problem.	
	analyze the algorithms in the context of efficiency.	
	apply the knowledge of stack and queue to design algorithm	
	Design application using sorting, searching and the concept of tree.	
UNITS	COURSE CONTENT	
	Introduction	(6L)
1	Basic Terminologies: Elementary Data Organizations, Data Structure Oper	ations: insertion, deletion, traversal etc.
-	Analysis of an Algorithm, Asymptotic Notations, Time-Space trade off.	
	Searching: Linear Search and Binary Search Techniques and their complexity	•
	Stacks and Queues	(8L)
	ADT Stack and its operations: Algorithms and their complexity analy	
2	Conversion and evaluation – corresponding algorithms and complexity anal	
	ADT, queue, Types of Queue: Simple Queue, Circular Queue, Priority Qu	ueue; Operations on each type of Queu
	Algorithms and their analysis.	
	Linked Lists	(6L)
	Singly linked lists: Representation in memory, Algorithms of several ope	erations: Traversing, Searching, Insertion
3	into, Deletion from linked list;	
5	Linked representation of Stack and Queue, Header nodes,	
	Doubly linked list: operations on it and algorithmic analysis;	
	Circular Linked Lists: all operations their algorithms and the complexity an	alysis.
	Trees	(6L)
	Basic Tree Terminologies, Different types of Trees: Binary Tree, Threader	d Binary Tree, Binary Search Tree, AV
4	Tree;	
	Tree operations on each of the trees and their algorithms with complexity and	
	Applications of Binary Trees. B Tree, B+ Tree: definitions, algorithms and	
	Graph	(6L)
5	Graph Terminology, Representation of graphs, Path Matrix, Graph Travers	sal, BFS, DFS, Minimum Spanning Tre
	Kruskal's Algorithm and Prim's Algorithm.	
	Sorting	(6L)
6	Objective and properties of different sorting algorithms: Selection Sort,	
	Merge Sort, Heap Sort; Performance and Comparison among all the method	
7	Hashing	(2L)
7	Review of Hashing, Hash Function, Collision Resolution Techniques	
	Addressing, Linear Probing, Quadratic Probing, Double Hashing, Rehashin	g, Extendible Hashing.
Reference l		1 C 1
	Fundamentals of Data Structures of C, Ellis Horowitz, Sartaj Sahni, Susan And	
	Python Programming A modular approach, Taneja Sheetal, Kumar Naveen Po	
	Data Structure and Algorithmic Thinking with Python, Narasimha Karumanch	-
	Data Structures and Algorithms in Python, Michael T. Goodrich, R.Tamassia,	witchael H. Goldwasser, wILEY
•	I annual r yunon, Jeeva Jose, Knanna r ubiisning Duthon Data Structures and Algorithms, Dr. Descent Accernal & Descentin Data	n Daalet
		a, I auni
•	Taming Python, Jeeva Jose, Khanna Publishing Python Data Structures and Algorithms, Dr. Basant Agarwal & Benjamin Bak Data Structures and Algorithms Using Python, Rance D. Necaise, Wiley	

Data Structures and Algorithms Using Python, Rance D. Necaise, Wiley

Code: MCA Contacts Ho	N-202 ours / Week: 4	Paper: Operating System Total Contact Hours: 40	Credit: 4	
<b>Course Outc</b>				
		course, students will be able to:		
		ents of OS and their working		
		ocess and thread and their scheduling policies		
		y management techniques		
		niques for managing memory, I/O, disk and files		
	lains the security and j	rotection features of an Operating System		
UNITS		COURSE CONTENT		
1	Structure of an OS Time Operating Sys	t of Operating systems, Systems, Types of Operating Systems, C - Layered, Monolithic, Microkernel Operating Systems, C ems, Distributed Operating Systems, Multiprocessor Opera ecture of Unix and Windows Operating Systems	Concept of Virtual Machine. Real	
2	communication, The CPU Scheduling: S SJF, RR, Priority, M processor schedulin Process Synchroni algorithm, Synchron Deadlocks: System Deadlock avoidance	eads: 7 state process model, Process scheduling, Operat eads overview, Benefits of threads, User and kernel threads cheduling criteria, Preemptive & non-preemptive schedulin fulti-level queue, Multi-level feedback queue), Comparative	s. ng, Scheduling algorithms (FCFS, ve study of the algorithms, Multi- solution – Peterson and Bakery hronization.	
3	and variable partitic Page allocation Har Virtual Memory: B fault, Working Set NRU and LRU), Th <b>Case Study:</b> Unix	cal and Physical address map, Memory allocation: Contig n– Internal and External fragmentation and Compaction; lware support for paging, Protection and sharing, Disadvan usics of Virtual Memory – Hardware and control structur Dirty page/Dirty bit – Demand paging, Page Replacement ashing /irtual Memory, Windows Virtual Memory	Paging: Principle of operation – tages of paging. res – Locality of reference, Page at algorithms(Optimal, FIFO, SC,	
4	structure, Allocation Grouping), Director PC Bus Structure, arbitration (Daisy c (Scheduling, Buffer	O Management mental File System Organization and Access Methods, methods (Contiguous, Linked, Indexed), Free-space man / Implementation (Linear list, Hash table), Efficiency and F I/O connections, Data transfer techniques (Programmed nain, Polling, Independent request), Blocking and non-blo ng, Caching, Spooling and device reservation, Error handli ile System, Windows File System	agement (Bit vector, Linked list, Performance. d, Interrupt driven, DMA), Bus ocking I/O, Kernel I/O subsystem	
5	Security and Prote Overview of Securi aspects of Security Control Matrix, Pro Case Study: Unix			
Reference Bo				
	berating System Conc tia Student Edition.	pts Essentials, 10th Edition by Avi Silberschatz, Peter C	Jalvin, Greg Gagne, Wiley	
		nals and Design Principles, 5th Edition, William Stallings,	Prentice Hall of India	
	Operating System Concepts, EktaWalia, Khanna Publishing House (AICTE Recommended Textbook – 2018)			
	Operating System: A Design-oriented Approach, 1st Edition by Charles Crowley, Irwin Publishing			
• Op	perating Systems: A M	odern Perspective, 2nd Edition by Gary J. Nutt, Addison- V	Vesley	
• De	sign of the Unix Oper	iting Systems, 8th Edition by Maurice Bach, Prentice-Hall	of India	
	· · · · · ·	Kernel, 3rd Edition, Daniel P. Bovet, Marco Cesati, O'Reil		

#### Master of Computer Application Code: MCAN-203 Paper: Object Oriented Programming with JAVA **Total Contact Hours: 40 Contacts Hours / Week: 4** Credit: 4 **Course Outcome:** After successful completion of this course, students will be able to: ✓ Use the characteristics of Java language in a program. Use variables and data types in program development. $\checkmark$ Identify and implement arrays, String and Selection Statements. $\checkmark$ Write Java programs using object-oriented programming techniques including classes, objects, methods, instance variables, interface. Design and implementation programs of Exception handling, Packages, Multithreading Programming, Window based programs. **COURSE CONTENT** UNITS (10L) **Object-Oriented Languages** Java's History, Creation of Java, Internet & Java, Byte-code, Its Features, Java Program Structure and Java's 1 Class Library, Data Types, Variables, and Operators, Operator Precedence; Selection Statements, Scope of Variable, Iterative Statement; Defining Classes & Methods, Creating Objects of a Class, Defining and Using a Class, Automatic Garbage Collection. Arrays and Strings: Arrays, Arrays of Characters, String Handling Using String Class, Operations on String Handling Using, String Buffer Class. **Classes and Inheritance** (10L) 2 Using Existing Classes, Class Inheritance, Choosing Base Class, Multiple Levels of Inheritance, Abstraction through Abstract Classes, Using Final Modifier, Packages: Understanding Packages, Defining a Package, Packaging up Your Classes, Adding Classes from a Package to Your Program, Understanding CLASSPATH, Standard Packages, Access Protection in Packages, Concept of Interface. Exception Handling: The concept of Exceptions, Types of Exceptions, Dealing with Exceptions, Exception Objects, Defining Your Own Exceptions. **Multithreading Programming** (10L)The Java Thread Model, Understanding Threads, The Main Thread, creating a Thread, Creating Multiple 3 Threads, Thread Priorities, Synchronization. Input / Output in Java: I/O Basic, Byte and Character Structures, I/O Classes, Reading Console Input, Writing Console Output, Reading and Writing on Files, Random Access Files, Storing and Retrieving Objects from File, Stream Benefits. Creating Applets in Java: Applet Basics, Applet Architecture, Applet Life Cycle, Simple Applet Display Methods, The HTML Applet Tag Passing Parameters to Applets. (10L) Working with Windows 4 AWT Classes, Window Fundamentals, Working with Frame, Creating a Frame Window in an Applet; Displaying Information within a Window. Working with Graphics and Texts: Working with Graphics, Working with Color, Setting the Paint Mode, Working with Fonts, Managing Text Output; Using Font Metrics, Exploring Text and Graphics, Working with AWT Controls, Layout Managers and Menus. **Reference Books:**

- The Complete Reference JAVA, Herbert Schildt, TMH Publication.
- JAVA and Object-Oriented Programming Paradigm, Debasish Jana, Prentice Hall of India
- Beginning JAVA, Ivor Horton, WROX Publication.
- JAVA 2 UNLEASHED, Tech Media Publications.
- JAVA 2(1.3) API Documentations.

Master of C	Computer Application
Code: MCA Contacts He	AN-204 Paper: Networking ours / Week: 4 Total Contact Hours: 40 Credit: 4
Course Ou	itcome:
After succe	essful completion of this course, students will be able to:
	nderstand the purpose of network layered models, network communication using the layered concept and able to ompare and contrast OSI and TCP/IP model.
✓ Di	ifferentiate among and discuss the four level of address (physical, logical, port and url) used by the internet TCP/IP otocols.
	nderstand the routing principals and algorithm such as distance vector routing and link state.
	dge the efficiency of the connection oriented and connectionless protocol.
	uniliar with the routing techniques, protocols and quality of service.
	xplain the concept of network security and cryptography.
UNITS	COURSE CONTENT
01115	Introduction (4L)
1	Direction of data flow (simplex, half duplex, full duplex), Network topology, categories of network (LAN, MAN WAN).
2	Protocol and Standard (4L)
2	Layered Task, The OSI model, TCP/IP protocol suite, Addressing
	Internetworking (10L)
	Internetworking concept, IPv4 and IPv6 Addressing, IPv4 protocol, IPv6 protocol, transition from IPV4 to IPV6
3	transition from IPv4 to IPv6, Address Mapping, Error Reporting, Multicasting, Unicast Routing Protocols, Distance
	Vector routing, Link state routing, Path vector routing, Multicasting Routing Protocols, Transmission Contro
	Protocol(TCP), User Datagram Protocol(UDP)
	Quality of Service (6L)
4	Data traffic, Congestion, congestion control, Quality of service, Techniques to improve QoS, Integrated services Differentiated service, QoS in Frame Relay, QoS in ATM
	DNS and Web (8L)
5	Name Space, Domain Name System, Distribution of Name Space, Remote Logging, Electronic Mail and File Transfer, WWW, Web document and HTTP, Network Management, Simple Network Management Protoco (SNMP)
	Network Security (8L)
6	Symmetric Key Cryptography, DES, AES, Asymmetric Key Cryptography, RSA, Diffie-Hellman, Security Services Digital Signature, Key Management, IP Security, SSL/TLS, PGP, Firewalls
Reference <b>B</b>	
• (	Computer Networks, Andrew S. Tanenbaum, Pearson Education, Fourth edition.
	Data and Computer Communication, William Stallings, Prentice hall, Seventh edition.
	High speed Networks and Internets, William Stallings, Pearson education, Second edition.
	Behrouz A Forouzan, - Data communication & Networking, TMH
	Behrouz A Forouzan, - TCP/IP Protocol Suite, TMH

Kelvin R Fall, W. Richard Stevens- TCP/IP Illustrated Volume 1, Addison Wesley

	'AN-E205APaper: Numerical and Statistical AnalysisHours / Week: 3Total Contact Hours: 30	Credit: 3
Course Ou	itcome:	
	essful completion of this course, students will be able to:	
	o understand approximation and propagation error.	
	o understand and implement different interpolation techniques.	
	o understand and implement integration techniques.	
✓ To	o understand and implement solutions for linear and algebraic and differential equations.	
UNITS	COURSE CONTENT	
	Approximation in numerical computation	(2L)
1	Truncation and rounding errors, Fixed and floating point arithmetic, Propagation of errors.	
	Interpolation	(3L)
2	Newton forward/backward interpolation, Lagrange's and Newton's divided difference Interpolation	(01)
	Numerical integration	(3L)
3	Trapezoidal rule, Simpson's1/3 rule, Romberg's Integration, Expression for corresponding error terms	
	Numerical solution of Linear equations	(3L)
4	Gauss elimination method, Matrix inversion, LU Factorization method, Gauss-Seidel iterative method.	
5	Numerical solution of Algebraic equation	(5L)
5	Bisection method, Regula-Falsi method, Newton-Raphson method, Iteration Method, Secant Method.	( )
	Numerical solution of ordinary differential equation	(4L)
6	Euler's method, Runge-Kutta methods, Taylor's series, method, Predictor Corrector methods and Fin	nite Differenc
	method.	
7	Least Square Curve fitting	(2L)
7	Linear & non-linear curve fitting	
	Introduction to Statistics& Probability	(8L)
8	Basic Statistics-measure of central tendency, dispersion.	
0	Probability, distribution introduction to mass function, density function, distribution function (Bind	omial, Poissor
	Normal).	
Reference <b>H</b>		
	R.S. Salaria: Computer Oriented Numerical Methods, Khanna Publishing House	
	C.Xavier: C Language and Numerical Methods.	
	Dutta & Jana: Introductory Numerical Analysis.	
	J.B.Scarborough: Numerical Mathematical Analysis.	
	Jain, Iyengar, & Jain: Numerical Methods (Problems and Solution).	
	Balagurusamy: Numerical Methods, Scitech. Baburam: Numerical Methods, Pearson Education.	

	CAN-E205B Hours / Week: 3	Paper: Computer Graphics Total Contact Hours: 30	Credit: 3
Course O			
After succ	essful completion of	this course, students will be able to:	
$\checkmark$	Identify the basic te	rminologies of Computer Graphics and interpret the mathematical foundation	n of the concepts of
	computer graphics.		-
✓ .	Apply mathematics	to draw basic primitives.	
✓	Illustrate the concept	ts of windowing and clipping and apply various algorithms to fill and clip poly	ygons.
✓ .	Understand and app	ly the core concepts of computer graphics, including transformation in two ar	nd three dimensions
	viewing and projecti	on.	
$\checkmark$	Create effective prog	grams using concepts of curves.	
✓ U	Inderstand the conce	pts of color models, lighting, shading models and hidden surface elimination.	
UNITS		COURSE CONTENT	
	Introduction		(6L)
	Overview of con	nputer graphics, representing pictures, preparing, presenting & interacting	g with pictures for
1		sualization & image processing; RGB color model, direct coding, lookup	
		Raster scan display, 3D viewing devices, Plotters, printers, digitizers, Light	
		levices; Computer graphics software.	1 /
	Graphics Primiti		(6L)
•		Circles as primitives, Scan conversion algorithms for primitives, Fill area	( )
2		filling, inside-outside test, Boundary and Flood-fill, Character generation, lin	
	attributes, charact		,
	2D Transformati	on and Viewing:	(6L)
	Transformations	(translation, rotation, scaling), matrix representation, homogeneous coo	rdinates, composit
3	transformations,	reflection and shearing, viewing pipeline and coordinates system,	window-to-viewpor
	transformation, c	lipping including point clipping, line clipping (Cohen-Sutherland, Lian,	g-Bersky), Polygoi
	clipping		
	3D Transformati	ons:	(4L)
4	Translation, rotati	on, scaling & other transformations. Rotation about an arbitrary axis in space	, Reflection through
	an arbitrary plane	; General parallel projection transformation; clipping, viewport clipping, 3D v	viewing.
	Curve:		(3L)
5		tion, surfaces, designs, Bezier curves, B-spline curves, end conditions fo	or periodic B-splin
	curves, rational B	-spline curves.	
	Hidden surfaces		(3L)
6		n, Z-buffer algorithm, Back face detection, BSP tree method, The Painter's	algorithm, scan-line
	algorithm; Hidder		
7	Color & shading		(2L)
		del; Interpolative Shading Models; Texture;	
eference			
		Baker, Computer Graphics, Pearson Eduction.	CT 1:
		<i>Jana</i> , Computer Graphics : Algorithms and Implementations, Pentice Hall	
		Feiner, J. Hughes, "Computer Graphics Principles and Practice", Pearson Ed	ucation.
•	D. Kogers, Procedu	al Elements for Computer Graphics, TataMcGraw-Hill Publications.	

D. Rogers, Procedural Elements for Computer Graphics, TataMcGraw-Hill Publications.
 D.Rogers, J.Adams, Mathematical Elements for Computer Graphics, TataMcGraw Hill Publication.

	CAN-E205CPaper: Probability and StatisticsHours / Week: 3Total Contact Hours: 30	Credit: 3
Course O		Crean: 5
	cessful completion of this course, students will be able to:	
	xplain the concept of probability and its feature in terms of random event, sample	space, favorable event.
	Describe the idea of random variable and the probability distribution.	· · · · · · · · · · · · · · · · · · ·
	alculate the expectation, standard deviation and moments.	
	ritically evaluate the underlying assumptions of analysis tools.	
	arry out basic statistical analysis of data.	
UNITS	COURSE CONTENT	
	Probability	(3L)
1	Sample Space, Probability Axioms, Combinatorics: probability of finite sa	
	probability and Bayes Theorem, Independence of Events,	
	Random Variables	(5L)
•	Discrete, continuous and mixed random variables, probability mass, prob	bability density and
2	cumulative distribution functions, mathematical expectation, moments,	2 2
	generating function, median and quantiles, Markov inequality, Chebyshev's inequality	. ,
	Special Distributions	(6L)
3	Discrete uniform, binomial, geometric, negative binomial, hypergeometric, Pois	son, continuous uniform
	exponential, gamma, Pareto, beta, normal	
	Joint Distributions	(3L)
4	Joint, marginal and conditional distributions, product moments, correlation and	regression, independence
	of random variables, bivariate normal distribution, problems.	
	Sampling Distributions	(2L)
5	The Central Limit Theorem, distributions of the sample mean and the samp	le variance for a norma
	population, Chi-Square, t and F distributions, problems.	
6	Descriptive Statistics	(2L)
0	Graphical representation, measures of locations and variability.	
	Estimation	(3L)
7	Unbiasedness, consistency, the method of moments and the method of maximu	um likelihood estimation
7	confidence intervals for parameters in one sample and two sample problem	s of normal populations
	confidence intervals for proportions, problems.	
	Testing of Hypotheses	(6L)
	Null and alternative hypotheses, the critical and acceptance regions, two types of	-
8	the most powerful test and Neyman-Pearson Fundamental Lemma, tests for on	· ·
	problems for normal populations, tests for proportions, Chi-square good	ness of fit test and its
	applications, problems.	
Reference		
	Vijay K. Rohatgi, A. K. Md. Ehsanes Saleh- An Introduction To probability And sta	atistics, John Wiley &Sons
	V.K.Rohatgi& A.K. Md. E. Saleh - An Introduction to Probability and Statistics J.S. Milton &J.C. Arnold- Introduction to Probability and Statistics	
	H.J. Larson -Introduction to Probability Theory and Statistical Inference.	
	S.M. Ross - Introduction to Probability and Statistics for Engineers and Scientists	

Code:MCA Contacts H	N-E205DPaper: Introduction to Cyber Securityours / Week: 3Total Contact Hours: 30Credit: 3
Course Ou	tcome:
After succe	ssful completion of this course, students will be able to:
✓ Kr	ow Fundamental knowledge in Cyber Security
✓ Ur	derstand the security challenges as well as the best practices that are essential to protect one from becoming the
vie	tims of cybercrimes.
	derstand the current status of cyber world.
	safe-guard the individual, society, organization and the government from the dangers of cyber frauds, scams,
	eats and attacks.
	le to further exploration in Cyber Security Domain.
UNITS	COURSE CONTENT
1	Introduction (2L
1	Introduction to Cyber Space, Information Systems, Need for Cyber Security
	Cyber Attacks (3L
2	Introduction to Cyber Attacks,
	Classification of Cyber Attacks,
	Classification of Malware, Threats
	Intrusion Detection and Prevention (2L
3	Vulnerability Assessment
	Intrusion Detection Systems
	Intrusion Prevention Systems
4	Authentication Methods (2L
4	Introduction to User Authentication Methods
	Biometric Authentication Methods, Biometric Systems
	Security Models (3L
5	Different Security Models and Security Mechanisms
	Information Security and Network Security
	Operating System Security (21
	Online Security (2L Web Security
6	Email Security
	Mobile Device Security, Cloud Security
	IoT & Social Media Security     (31)
	IoT Security
7	Cyber Physical System Security
	Social Media Security
	Security and Virtual Currency (31
	Virtual Currency
8	Block Chain Technology
	Security Auditing
	Cyber Crimes (4L
9	Introduction, Different Types of Cyber Crimes, Scams and Frauds, Analysis of Crimes, Human Behavio
	Stylometry, Incident Handling, Investigation Methods, Criminal Profiling, Cyber Trails
	Digital Forensics (3L)
10	Digital Forensics, History, Challenges, Branches of Digital Forensics, Digital Forensic Investigation Method
	Reporting, Management of Evidence
	Cyber Law (3L
11	Cyber laws, Cyber terrorism, Information Technology Act 2000 and Amendments, Evidentiary value of
11	Email/SMS, Cybercrimes and Offenses dealt with IPC, RBI Act and IPR Act in India, Jurisdiction of Cyber
	Crime, Cyber Security Awareness Tips
eference B	
	undamentals of Cyber Security By MayankBhushan, BPB Publications
	ttps://heimdalsecurity.com/pdf/cyber_security_for_beginners_ebook.pdf
	nformation Security & Cyber Laws, Gupta & Gupta, Khanna Publishing House
	Certified Ethical Hacker Certification Exam by William Manning
• L	Data communication and Networking by Behrouz A. Forouzan, McGraw Hill Education (India) Pvt. Ltd. http://larose.staff.ub.ac.id/files/2011/12/Cyber-Criminology-Exploring-Internet-Crimes-and-Criminal-Behavior.pdf

Code:MCAN Contacts Ho	N-E205E ours / Week: 3	Paper: Introduction to IoT Total Contact Hours: 30	Credit: 3
<ul> <li>✓ Expla</li> <li>✓ Descr</li> <li>✓ Unde</li> <li>✓ Expla</li> </ul>	ssful completion ain what Internet o ribe key technolog rstand wireless se ain resource mana	of this course, students will be able to: f Thins is ies in Internet of Things. noor network architecture and its framework along w gement in the Internet of Things. models for the Internet of Things.	ith WSN applications.
UNITS		COURSE CONTENT	
1	(6) What is the Internet of Things? : History of IoT, About IoT, Overview and Motivations, Examples of Applications, Internet of Things Definitions and Frameworks : IoT Definitions, IoT Architecture, General Observations, ITU-T Views, Working Definition, IoT Frameworks, Basic Nodal Capabilities		
2	Fundamental IoT Mechanisms And Key Technologies       (61         Identification of IoT Objects and Services, Structural Aspects of the IoT, Environment Characteristics, Traffic Characteristics, Scalability, Interoperability, Security and Privacy, Open Architecture, Key IoT Technologies, Device Intelligence, Communication Capabilities, Mobility Support, Device Power, Sensor Technology, RFID Technology, Satellite Technology,		
3	RFID: Introdu Framework: E Things. Wirele Networking N	ncy Identification Technology ction, Principle of RFID, Components of an RFI PCIS & ONS, Design issues, Technological challeng ess Sensor Networks: History and context, WSN odes, Securing Communication WSN specific Io Various integration approaches, Data link layer pro-	ges, Security challenges, IP for IoT, Web of Architecture, the node, Connecting nodes, $\Gamma$ applications, challenges: Security, QoS,
4	Resource Management In The Internet Of Things           Clustering, Software Agents, Clustering Principles in an Internet of Things, Architecture, Design Guand Software Agents for Object Representation, Data Synchronization. Identity portrayal, management, various identity management models: Local, Network, Federated and global web identicentric identity management, device centric identity management and hybrid-identity management and trust.		chronization. Identity portrayal, Identity k, Federated and global web identity, user-
5			
CF Int Ya Pu Da Ne (C	thuru Raj and An RC Press, Taylor a ternet of Things asuura, H., Kyun iblishing. avid Hanes, Gon etworking Techno isco Press Indian	A Hands-on Approach, Arshdeep Bahga and Vijay M g, C.M., Liu, Y., Lin, YL., Smart Sensors at zalo Salgueiro, Patrick Grossetete, Robert Bart logies, Protocols, and Use Cases for the Internet of	Iadisetti, Universities Press.the IoT Frontier, Springer Internationalon, Jerome Henry, IoT Fundamentals:

	CAN-E205F Hours / Week: 3	Paper: Automata Theory & Computational Complexity Total Contact Hours: 30	Credit: 3
Course O			
		f this course, students will be able to:	
		al notation for strings, languages and machines.	
✓ D	Design and Implement	nt Finite automata to accept a string of a language.	
✓ F	for a given language	determine whether the given language is regular or not.	
		rammars to generate strings of context free language.	
✓ D	Determine equivalen	ce of languages accepted by Push Down Automata and languages generated by	y context free
	rammars		
		ze the hierarchy of formal languages, grammars and machines.	
✓ D	Distinguish between	computability and non-computability and Decidability and un-decidability.	
UNITS		COURSE CONTENT	
1	Introduction		(2L)
1	Introduction: Alp	habet, languages and grammars, productions and derivation, Chomsky hierarc	hy of languages.
		ges and finite automata:	(6L)
		ons and languages, deterministic finite automata (DFA) and equiva	alence with regula
2		deterministic finite automata (NFA) and equivalence with DFA, regular	
		quivalence with finite automata, properties of regular languages, pumping	g lemma
		ages, minimization of finite automata)	
		iguages and pushdown automata:	(6L)
3		nmars (CFG) and languages (CFL), Chomsky and Greibach normal fo	
5		ata (PDA) and equivalence with CFG, parse trees, ambiguity in CFG,	pumping lemma for
		uages, deterministic push down automata, closure properties of CFLs.	
	Turing machine		(8L)
		for Turing machines (TM), Turing recognizable(recursively enumerable)	
4		ages and their closure properties, variants of Turing machines, nond	
		h deterministic TMs, unrestricted grammars and equivalence with	
	and context sensit	rs.Context Sensitive Language, The model of Linear Bounded Automaton, r	relation between LB.
	Decidability:	ive language	(4L)
5		dable language and undecidable language, Halting problem of TM, Halting pr	
	Complexity	dable language and underdable language, flatting problem of five, flatting pr	(4L)
6		unctions, The classes P and NP, Polynomial time reduction and NP comp	
0		theorem, Church-Turing Thesis	Jieteness, 5711 15 14
Reference			
٠		tomata Theory, Languages, and Computation, John E. Hopcroft, Rajeev Molucation Asia.	otwani and Jeffrey D
		cory of Computation, Harry R. Lewis and Christos H. Papadimitriou, Pearson I	Education Asia.
•	Theory of Compute	r Science, Automata Languages and computation, Mishra and Chandra shekar	an, 2nd edition, PHI
•	Automata and Com	putability, Dexter C. Kozen, Undergraduate Texts in Computer Science, Sprin	
		Theory of Computation, Michael Sipser, PWS Publishing.	
•	Introduction to Lan	guages and The Theory of Computation, John Martin, TataMcGraw Hill., PEA	RSON.

Master o	of Computer Application	
Code: MC Contacts I	CAN-291Paper: Data Structure Lab with PythonHours / Week: 4Total Contact Hours: 40Credit: 2	
Course Ou	utcome:	
	essful completion of this course, students will be able to:	
✓ Te	o understand linear and non-linear data structures.	
	o understand different types of sorting and searching techniques.	
✓ Te	o know how to create an application specific data structure.	
✓ Te	o solve the faults / errors that may appear due to wrong choice of data structure.	
✓ T	o analyze reliability of different data structures in solving different problems.	
UNITS	COURSE CONTENT	
1	Implementation of data structure operations (Insertion, deletion, traversing, searching) on array. Linear sear Binary search.	rch,
2	Implementation of stack, queue operation using array. Pop, Push, Insertion, deletion, Implementation of circu queue. Infix to postfix conversion, postfix expression evaluation	ular
3	Implementation of linked lists: Single linked list, circular linked list, double linked list, doubly circular linked limplementation of stack and queue using linked list. Merging two linked list, Linked list representation of polynomial, polynomial addition, polynomial multiplication.	of a
4	Tree: creating Binary Search tree, recursive and non-recursive traversal of BST, deletion in BST, calculating hei of a BST, building AVL tree.	ight
5	Implementation of sorting techniques: selection, bubble, quick sort, insertion sort, merge sort, heap implementation of priority queue. Hash table implementation.	sot,
6	Implementation of Graph: representation, searching, BFS, DFS	

Code: MC		Paper: Operating System Lab (Unix)	
	Hours / Week: 4	Total Contact Hours: 40	Credit: 2
Course O			
		his course, students will be able to:	
		UNIX Commands from the command line, and create Shell Scri	pts to customize their UNIX
	Vorking Environment.		
		heir processes within UNIX through system calls.	
		heir files within the UNIX through system calls. or handling asynchronous events through signals (Software Interru	
		ocess communication using FIFOs, Message Queues, Semaphores, nming to design Client-Server Environment.	and Shared Memory.
		nent Multithreaded Programming Environment.	
		COURSE CONTENT	
UNITS	Shell programmin		
1		g naking a script executable, shell syntax (variables, Conditions, co	ontrol structures functions and
1	commands).	naking a script executable, shen syntax (variables, conditions, c	sinter structures, runctions and
	Process		
2		ss, replacing a process image, duplicating a process image, waiting	for a process. Zombie Process.
_	Orphan Process		·····, _····,
2	File Handling		
3	0	les (use create(), open(), read(), write(), close(), lseek(), dup()).	
4	Signal		
		locking, Suspending, Delivering Signals, Various Signal Related F	unctions.
	Inter-process com		
		s pipe(), popen(), pclose()),	
5		Os, accessing FIFO),	
5		se functions msgget(), msgsnd(), msgrcv(), msgctl()),	
		unctions semctl(), semget(), semop())	
	• •	se functions shmget(), shmat(), shmdt(), shmctl())	
6	Sockets:		
0		Sockets, Socket Options, Cliient /Server Example, Name and Add	ress Conversions
_	POSIX Threads		
7		pthread functions (viz. pthread_create(), pthread_join(), pthr	read_exit(), pthread_attr_init(),
	pthread_cancel())		

#### Master of Computer Application

# Code: MCAN-293Paper: Object Oriented Programming Lab using JAVAContacts Hours / Week: 4Total Contact Hours: 40Credit: 2

#### **Course Outcome:**

After successful completion of this course, students will be able to:

- ✓ Apply object-oriented principles or features in software design process to develop Java programs for real life applications.
- $\checkmark$  Reduce the complexity of procedural language by employing different OOP technologies for developing robust and
- reusable software.
  ✓ Develop programs using stream classes for various I/O operations and design concurrent.
- ✓ Design graphical user interface to develop user interactive applications.

UNITS	COURSE CONTENT	
1	1 Assignments on class, constructor, overloading, inheritance, overriding.	
2	2 Assignments on wrapper class, arrays.	
3	Assignments on developing interfaces- multiple inheritance, extending interfaces.	
4	Assignments on creating and accessing packages.	
5	Assignments on multithreaded programming	
6	Assignments on applet programming	



# **MASTER OF COMPUTER APPLICATION**

Syllabus w.e.f. the Academic Session 2020-2021





MAULANA ABUL KALAM AZAD UNIVERSITY OF TECHNOLOGY WEST BENGAL

### Second Year: Semester-III

Code: MC	AN-301 Paper: Software Engineering using UML	
	Iours / Week: 4 Total Contact Hours: 40	Credit: 4
Course Ou		
After succe	ssful completion of this course, students will be able to:	
🗸 Ar	halyze the problem scenario and identify classes/ objects and their properties, relationship in class	s model.
✓ De	emonstrate the conceptual modeling techniques of UML for solving Real-World problem.	
✓ To	learn software development life cycle for Object-Oriented solutions for Real-World Problems.	
	bility to apply the concepts of object oriented methodologies to analyze requirements and design	to the point where it
	ready for implementation.	
	emonstrate the concept of Testing to measure quality of software.	
UNITS	COURSE CONTENT	
	Introduction to Software Engineering:	(2L)
1	What is Software Engineering? Software Engineering Concepts, Software Engineering Deve	elopment Activities,
	Managing Software Development.	
	Object Oriented Concept and Modelling:	(5L)
	<b>Object-Oriented Principals and Concepts:</b> Classes and Object, Modularity, Abstraction	
2	Object Relationship like Association, Aggregation and Composition; Inheritance, Polymorp	phism and Dynamic
2	Binding Interfaces Model: Importance of Modeling, Object Oriented Modeling	
	<b>Identifying the Elements of an Object Model:</b> Identifying classes and objects, Specific	wing the attributes
	Defining operations, Finalizing the object definition.	ying the attributes
	Introduction to UML:	(3L)
3	Overview of UML, Conceptual Model of UML, Architecture, S/W Development Life Cycle.	(51)
	Basic and Advanced Structural Modeling:	(7L)
4	Classes Relationship, Common mechanism, Diagrams, Class Diagram, Advanced	
	Relationship, Interface, Types and Roles, Packages, Object Diagram.	,
	Basic and Advanced Behavioral Modeling:	(7L)
5	Interactions, Use cases, Use Case Diagram, Sequence Diagram, Collaboration Diagram, Ir	nteraction Diagram,
	Activity Diagram, State Chart Diagram.	
6	Architectural Modeling:	( <b>3</b> L)
Ű	Artifacts, Artifact Diagram, Implementation Diagram, Deployment Diagram.	
	Object-Oriented Design:	(5L)
7	Generic components of Object-Oriented Design model, System Design process, Partitioning t	
	Concurrency and subsystem Allocation, Task Management component, Data Management Co	omponent, Resource
	Management Component, Inter Sub-system Communication.	(41)
8	<b>Object Oriented Analysis:</b> Iterative Development, Unified process & its Phases: Inception, Elaboration, Constr	(4L)
0	Understanding requirements.	uction, Transition,
	Object Oriented Testing:	(4L)
9	Overview of Testing and object oriented Testing, Types of Testing, Object oriented Testing s	
2	design for Object-Oriented software, Inter class test case design.	
Reference B		
	The Unified Modeling Language User Guide, Grady Booch, James Raumbaugh, Ivar Jacobson.	
• (	Dbject Oriented Software Engineering, Ivar Jacobson, ACM Press	
	Applying UML and Patterns, Craig LarmanMotilalUk Books Of India	
	Object-Oriented Software Engineering: Using UML, Patterns, and Java, Bernd Bruegge, Allen Du	
• 5	Software Engineering – A Practitioner's Approach, Roger. S. Pressman and Bruce R. Maxim, Mc	Graw Hill

Master of Computer Application Code: MCAN-302 Paper: Artificial Intelligence Contacts Hours / Week: 4 **Total Contact Hours: 40** Credit: 4 **Course Outcome:** ✓ After successful completion of this course, students will be able to understand the underlying assumption of philosophy of the logical sequences of real life problem by applying State Space Search behind the limitation of non-solving method of conventional computational approach. Incorporating heuristic search technique on Game Playing. Various strategies of representing knowledge with decision making algorithms. Creation of substantial domain knowledge base with meta data. Application of knowledge representation issues using Prolog/LISP. To recognize the adoption of new system through learning by an Intelligent System and processing of Natural Language. Ability to apply machine learning techniques to solve real world problems and how Expert Systems can be carried out by the help of learning, analyzing by applying various search techniques and resolute to provide solutions. UNITS **COURSE CONTENT Introduction to Intelligent Systems:** (8L) 1 Overview of Artificial intelligence- Problems of AI, AI technique, Tic - Tac - Toeproblem. **Search Techniques:** (10L) Problems, Problem Space & search. 2 Heuristic Search Techniques, Game planning -Minimax search procedure, adding alpha beta cut-off's, Iterative Deepening. (7L) **Knowledge Representation Issues:** Representing knowledge using rules. Weak slot & filler structures. 3 Strong slot & filler structures. Implementation of Knowledge with Prolog Programs. Basic knowledge of programming language like Prolog & Lisp. Adoption of New Knowledge: (10L) Deep Learning: Introduction to Neural Networks, Convolution of New Knowledge 4 Natural language processing, Understanding. Learning – induction & explanation based learning. **Expert systems:** (5L) 5 Expert system shells, knowledge acquisition. **Reference Books:** Artificial Intelligence: A Modern Approach, Stuart Russell & Peter Norvig, Pearson Education. Artificial Intelligence, Rich & Knight, TMH. Reference Books Artificial Intelligence & Intelligent Systems, N.P Padhy, Oxford University Press. Introduction to Artificial Intelligence & Expert Systems, Dan W. Patterson, PHI. Artificial Intelligence: A new Synthesis, Nils J. Nilsson, Morgan Kaufmann Publishers, Inc.

• M.C. Trivedi, Artificial Intelligence, Khanna Publishing House, New Delhi

Code: MC Contacts F	AN-303 Iours / Week: 4	Paper: Design and Analysis of Algorithm Total Contact Hours: 40	Credit: 4
Course Ou			
		this course, students will be able to:	
✓ U		yze the running times of algorithms based on asymptotic analysis and	d justify the correctness of
		nd-conquer paradigm and explain when an algorithmic design situation	n calls for it.
		ment the greedy paradigm for a given problem.	
		rogramming paradigm and implement it.	
		ment the Back Tracking and Branch-&-Bound problem.	
✓ Fo		ngineering problem model it using graph and write the correspondi	ing algorithm to solve the
1		nalyze randomized algorithms (expected running time, probability of e	error).
UNITS		COURSE CONTENT	)
1	average and wors	f algorithm. Analysis of algorithm: Asymptotic analysis of co st-case behavior; Performance measurements of Algorithm, Time rsive algorithms through recurrence relations: Substitution method, Re	e and space trade-offs,
2	Divide & Conque Strassen's matrix	General Method -knapsack problem - Tree vertex splitting - Job sec	
3	Dynamic Program Assembly-line pro	mming: ogramming, Matrix Chain Multiplication, 0-1 knapsack problem	(6L
4	<b>Graph Algorithn</b> Introduction to Sp	<b>ns:</b> banning tree, growing a minimum spanning tree, Prims and Kruskal Al	(4L) lgorithm
5	Back Tracking a Back Tracking: (	nd Branch-&-Bound: General Method – 8-queens - Sum of subsets - Graph Coloring –Ha eral Method - Traveling Salesperson problem.	(8L
6	Lower Bound Th Comparison trees Hard and NP-Con	- Oracles and advisory arguments - Lower bounds through reduction	(6L n - Basic Concepts of NP-
eference B	looks:		

 A.V. Aho, J.E. Hopcroft, J.D. Ullmann, 1974, The Design and Analysis of Computer Algorithms, Addison Wesley, Boston.

	AN-E304A Paper: Image Processing
Contacts H Course Ou	Iours / Week: 3         Total Contact Hours: 30         Credit: 3
	ssful completion of this course, students will be able to:
	escribe the fundamental concept of the digital image processing system.
	periment the images in the frequency domain and spatial domain using various transforms.
	aluate the techniques for image enhancement and restoration.
	plain different feature extraction techniques for image analysis and recognition.
	tegorize various compression techniques.
	evelop any image processing application.
UNITS	COURSE CONTENT
	Introduction (4L
1	Background, Digital Image Representation, Fundamental steps in Image Processing, Elements of Digital Image
	Processing - Image Acquisition, Storage, Processing, Communication, Display.
	Digital Image Formation (4L
2	A Simple Image Model, Geometric Model- Basic Transformation (Translation, Scaling, Rotation), Perspective
	Projection, Sampling & Quantization - Uniform & Non uniform.
	Mathematical Preliminaries (61
2	Neighbour of pixels, Connectivity, Relations, Equivalence & Transitive Closure; Distance Measures,
3	Arithmetic/Logic Operations, Fourier Transformation, Properties of The Two Dimensional Fourier Transform,
	Discrete Fourier Transform, Discrete Cosine & Sine Transform
	Image Enhancement (6L
	Spatial Domain Method, Frequency Domain Method, Contrast Enhancement -Linear & Nonlinear Stretching,
4	Histogram Processing; Smoothing - Image Averaging, Mean Filter, Low-pass Filtering; Image Sharpening. High-
	pass Filtering, High-boost Filtering, Derivative Filtering, Homomorphic Filtering; Enhancement in the frequency
	domain - Low pass filtering, High pass filtering.
	Image Restoration (5L
5	Degradation Model, Discrete Formulation, Algebraic Approach to Restoration - Unconstrained & Constrained;
	Constrained Least Square Restoration, Restoration by Homomorphic Filtering, Geometric Transformation – Spatial Transformation, Gray Level Interpolation.
	Image Segmentation (5L
	Point Detection, Line Detection, Edge detection, Combined detection, Edge Linking & Boundary Detection –
6	Local Processing, Global Processing via The Hough Transform; Thresholding - Foundation, Simple Global
Ũ	Thresholding, Optimal Thresholding; Region Oriented Segmentation - Basic Formulation, Region Growing by
	Pixel Aggregation, Region Splitting & Merging.
Reference <b>B</b>	
	Digital Image Processing, Rafael C.Gonzalez& Richard E.Woods, Pearson
	Fundamentals of Digital Image Processing, Anil K. Jain, Pearson Education-2003.
	Digital Image Processing, Jahne, Springer India
• I	Digital Image Processing & Analysis, Chanda & Majumder, PHI

Digital Image Processing & Analysis, Chanda & Majumder, PHI
Fundamentals of Digital Image Processing, Jain, PHI

	CAN-E304B Hours / Week: 3	Paper: Web Enabled JAVA Programming Total Contact Hours: 30	Credit: 3
Course O		Total Contact Hours. 50	Cituit. 5
		course, students will be able to:	
		ing methodology of JSP, servlet and JSF Frameworks	
		cation using JSP and servlet and database.	
	esign and develop a Web		
		plying concepts and error handling techniques.	
JNITS		COURSE CONTENT	
	Core Java Overview:		(4L
1	The Java.sql package, Execution of SQL stat Transaction Methods. I	tts, Exception Handling, Multi Threading Introduction to JDE JDBC Drivers, Executing SQL commands using JDBC I tements, Execution of Stored Procedures using JDBC. Intro Introduction to JNDI, Introduction to Data Source and Connec Servers Overview of J2EE Technologies. (6L)	Drivers, static and dynami duction to Transactions an
	Introduction to Java S	• • • •	(6L
2	Deploying a Servlet,	ntents, Servlet life Cycle and Life cycle methods, Servlet Re Servlet State Transitions, Servlet Config and Servlet Conte let Synchronization and Thread Model. Maintaining Client Sta sion Tracking. (8L)	equest and Response Mode ext, Servlet Redirection an
	Introduction to JSP :		(61
3	Syntax, JSP syntactic expressions, scriptlets,	Components, Servlets vs. JSP, JSP Lifecycle, JSP Page Lifecy elements, JSP element syntax, Template content. JSP elem actions. JSP Standard Actions: jsp:useBean, jsp:getPreoperty, jsp:param,java Server Pages Standard Tag Library(JSTL).	ents-directives, declaration
	Introduction to JSF F	rameworks:	(8L
4	Beans: A Sample Ap Navigation, Standard JJ Process, Using Standard Event Listener Tags, I Converters and Validar Component Developer Converters, Implement Handlers in JSF 1.1.	bele Example, Sample Application Analysis, Development Env oplication, Bean Scopes Configuring Beans, Navigation, S SF tags, Data tables, conversion and validation Overview of the d Converters. Event Handling: Life Cycle Events, Value Ch immediate Components, Passing Data from the UI to the S tors: Classes for Implementing Custom components, Tags an 's Toolbox, Encoding: Generating Markup, Decoding: Proces- ting Custom Component Tags, The TLD File, The Tag H	Static Navigation, Dynami the Conversion and Validatio ange Events, Action Events erver, Custom Components d Components, The Custor ssing Request Values, Usin fandler Class, Defining Ta
	AJAX:		(6L
5	Completion. Realtime Hybrid Components, Code,Ajax4jsf,Impleme Ajax4jsf.Introduction to	vaScript Libraries, The Prototype Library, The Fade Anythin Validation, Propagating Client-Side View State Direct Web R Keeping JavaScript Out of Renderers, Transmitting JSP T enting Form Completion with Ajax4jsf,Implementing o Java Web Services.	Remoting, Ajax Component ag Attributes to JavaScrip
eference			
	publication, 2007.	<ul> <li>Programming- J2EE 1.3 Edition- SubrahmanyamAllamaraj</li> <li>econd Edition-David Geary, CayHorstmann-Prentice Hall-2007</li> </ul>	-

Code:MCAN Contacts Ho	1 1 8	redit: 3
Course Outc		
	ful completion of this course, students will be able to:	
🗸 Unde	erstandand identify the architecture and infrastructure of cloud computing, including SaaS, PaaS, IaaS, p	ublic
	l, private cloud, hybrid cloud.	
	ribe the core issues of cloud computing such as security, privacy, and interoperability to choose the appr	opriate
	ologies, algorithms, and approaches for the identified problems.	
	yze various cloud computing solutions.	
	erstand cloud Storage systems and Cloud security, the risks involved, its impact. y knowledge for solving real life cloud computing problem scenario and illustrate solutions.	
UNITS	COURSE CONTENT	
UNITS	Basics of Cloud Computing	[4]
	Defining a Cloud, Cloud Types – NIST Cloud Reference Model, Cloud Cube Model, Deploymen	
1	(Public, Private, Hybrid and Community Clouds), Service Models - IaaS, PaaS, SaaS, Ben	
	Advantages of Cloud Computing	
2	Concepts of Abstraction and Virtualization	[4]
2	Taxonomy of Virtualization, Reference model for Virtualization	
	Services and Applications by Type	[5]
	IaaS – Basic Concept, Workload, Partitioning of Virtual Private Server Instances, Pods, Aggregations	, Silos
3	PaaS – Basic Concept, Tools and Development Environment with examples	
	SaaS - Basic Concept and Characteristics, Open SaaS, examples of SaaS Platform	
	Identity as a Service (IDaaS), Compliance as a Service (CaaS)	
	Concepts of Service Oriented Architecture (SOA) and Web Service (WS)	[2]
4	Service Oriented Architecture - Basics, Terminologies, Components, Standards and Technologies, Be	enefits an
-	Challenges	
	Web Services – Basics, Characteristics, Terminologies, Characteristics and Scope, Business Models	
5	Cloud-based Storage	[3]
	Cloud File Systems, including GFS and HDFS Cloud Security	[2]
	Cloud security concerns, security boundary, security service boundary	[2]
	Overview of security mapping	
6	Security of data: cloud storage access, storage location, tenancy, encryption, auditing, compliance	
	Identity management (awareness of identity protocol standards)	
	Risk Management and Compliance	
	Cloud Security	[2]
	Cloud security concerns, security boundary, security service boundary	
	Overview of security mapping	
7	Security of data: cloud storage access, storage location, tenancy, encryption, auditing, compliance	
	Identity management (awareness of identity protocol standards)	
	Risk Management and Compliance	
	Introduction to Various Web Services	[6]
8	Amazon Web Services, Google Web Services, Microsoft Cloud Services	[0]
9	Cloud Federation	[2]
	Definition, different scenario description, replace ability and negotiation mechanism	
eference Boo		vaati -
	stering Cloud Computing by RajkumarBuyya, Christian Vecchiola, S. ThamaraiSelvi, McGraw Hill Edu ud Computing Bible by Barrie Sosinsky, Wiley India Pvt. Ltd	ication
	ud Computing Blole by Barrie Sosinsky, whey India PVI. Etd ud Computing: A Practical Approach by Anthony T. Velte, Tata Mcgraw-Hill	
	lding Applications in Cloud: Concept, Patterns and Projects by Moyer, Pearson.	
	ud Security by Ronald Krutz and Russell Dean Vines, Wiley-India	

Code: MCAN- Contacts Hours		Paper: Web Technology using PHP Total Contact Hours: 30	Credit: 3
<b>Course Outco</b>			
<ul> <li>✓ After s variable</li> <li>✓ Incorpo</li> <li>✓ Implen</li> </ul>	uccessful complet es, constants, oper orating HTML forn nentation of Decisi	ion, Loops, Functions, Array and Exception Handling concepts using I	GET Method.
	y to connect with I		
•		on using JavaScript & JQuery.	
	ting Forms using		
UNITS	<b>T</b> ( <b>1</b> ( <b>1</b> )	COURSE CONTENT	(41)
1	Evaluation of PH	<b>Web Technology &amp; implementation of PHP Programs:</b> HP. Basic Syntax. Defining variables and constants. PHP Data type L: Form Creation, Handling of Forms, Submission of Forms. POST& O	
2		<b>Form With PHP</b> . Data Dealing with Multi-value files. Generating File uploaded form	(4L). Redirecting a form afte
3	Making Decision What is a function Creating and ac function. PHP Array. Created array. Looping we Understanding E	etions, String, Array & Exception Handling ns. Doing Repetitive task with looping. Mixing Decisions and looping on? Cookies, Session and in-built functions. ccessing String. Searching & Replacing String. Formatting String ating index based and Associative array. Accessing array Element. I with associative array using each() and foreach(). Some useful Library Exception and error. Try, catch, throw.	g. String Related Librar
4	Introduction to R	ectivity with MySql RDBMS. Connection with MySql Database. Performing basic databas Select). Setting query parameter. Executing query Join (Cross joins	
5		<b>Query</b> Javascript. Three ways to use Javascript. Working with event Query. Validation using JQuery. JQuery Forms. JQuery Examples.	(4L) ts Client-side Validation
6 Connecting Forms using AJAX Concept Introduction to AJAX. PHP with AJAX. Working with database.		(4L)	
Reference Bool	<b>KS:</b>		
• The Jo	y of PHP Program	mming: A Beginner's Guide to Programming Interactive Web App	olications with PHP and

 Open Source for the Enterprise: Managing Risks, Reaping Rewards, DanWoods and GautamGuliani, O'Reilly, Shroff Publishers and Distributors, 2005.

	CAN-E304E Hours / Week: 3	Paper: Android Application Development Total Contact Hours: 30	Credit: 3	
Course O				
		this course, students will be able to:		
		plication development trends and Android platform		
✓ A	analyze the need of si	mple applications, game development, Location map based services		
		, email, service, binding and deploying APks		
🗸 T	o develop, deploy an	d maintain the Android Applications.		
UNITS		COURSE CONTENT		
	Android Fundam	entals	(6L)	
1	Mobile Application	n development and trends - Android overview and Versions - Android	d open stack, features -	
1	Setting up Android	d environment (Eclipse, SDK, AVD)- Simple Android application deve	elopment – Anatomy of	
	Android applicatio	ns – Activity and Life cycle – Intents, services and Content Providers		
	Android User Inte	erface	(6L	
	Layouts: Linear, Absolute, Table, Relative, Frame, Scroll view, Resize and reposition - Screen orientation - Views:			
2	Text view, Edit Text, Button, Image Button, Checkbox, Toggle Button, Radio Button, Radio Group, Progress Bar,			
		Auto complete Text, Picker, List views and Web view- Displaying pictures with views: Gallery and Image View,		
		rid view – Displaying Menus: Helper methods, Option and Context		
	Data Persistence		(6L)	
3	Shared User preferences - File Handling: File system, System partition, SD card partition, user partition, security,			
		nal Storage - Managing data using SQLite -User defined content provide		
		orking And Services	(6L)	
4		Sending and Receiving - Sending email and networking - Downloading b		
		vices - Developing android services: create your own services, performin	ig long running task in a	
		g repeated task in a service		
-		And Publish Android Application	(6L	
5		vices: Display map, zoom control, view and change, Marking, Geocodin	g, Get location - Publish	
		ns and Deployment		
Reference		$\mathbf{A}_{\mathrm{res}} = \mathbf{D}_{\mathrm{res}} = \mathbf{D}_{\mathrm{res}} = \mathbf{M}_{\mathrm{res}} + \mathbf{M}_{\mathrm{res}} + \mathbf{M}_{\mathrm{res}} = \mathbf{M}_{\mathrm{res}} + \mathbf{M}_{\mathrm{res}$	W.1	
		Application Development, WeiMeng Lee, (2012) Wrox Publications (John valuating Geogle's Mahila Development Platform Ed Purpetta (2010) Th		
•	3rd edition, North Ca	ducing Google's Mobile Development Platform, Ed Burnette (2010), Tharolina USA	le riaginatic rubiishers,	
		1 4 Application Development, Reto Meier (2012), Wrox Publications (Joh	wilow Now Vork)	

Professional Android 4 Application Development, Reto Meier (2012), Wrox Publications (John Wiley, New York). Programming Android: Java Programming for the New Generation of Mobile Devices, Zigurd Mednieks, Laird Dornin, Blake Meike G, Masumi Nakamura (2011), OReilly Media, USA

	urs / Week: 3 Total Contact Hours: 30 Credit	t: 3
Course Outc		
	ful completion of this course, students will be able to:	
	erstand the fundamental knowledge of Data Science and the task of Data Science people.	
✓ Und	erstand fundamental of statistics.	
✓ Calo	sulate the correlation, covariance, central tendency.	
🗸 Esti	mate confidence interval.	
✓ Perf	orm hypothesis testing.	
	erstand the mechanics of regression analysis.	
	y out regression, classification using kNN, decision tree.	
✓ Use	clustering method to cluster records.	
UNITS	COURSE CONTENT	
	Introduction to Data Science	(2)
1	Define Data Science, why data science, data science in business	Ì
2	Descriptive Statistics	(4)
2	Matrix, Matrix operations, Sample, Population, Descriptive statistics, Central tendency, outlier detection	
3	Inferential Statistics	(4]
5	Basics of probability, probability distribution, Central Limit theorem	
	Hypothesis testing	(6]
4	Null and Alternate Hypothesis, Making a Decision, and Critical Value Method, p-Value Method and Type	es of
	Errors, Two-Sample Mean and Proportion Test	( 4)
5	Regression Analysis	(4]
3	Fundamentals of Regression analysis, assumption of regression analysis, accuracy, validity, Dealing with categorical data	
	Classification	(4)
6	Introduction, Logistic regression, model building and evaluation	(4)
	Clustering	(2)
7	Introduction to clustering, k-means clustering, hierarchical clustering	(2)
	Decision tree and kNN	(4)
8	Introduction to decision tree, regression tree, truncation & pruning, random forest, kNN for regression,	(
	classification, weighted kNN	
eference Bo		
	roducing Data Science; Davy Cielen, Arno D Meysman and Mohamed Ali; Dreamtech Press	
	etical Statistics for Data Scientists; Peter Bruce and Andrew Bruce; O"Reilly Media Inc.	
<ul> <li>Do</li> <li>Mi</li> </ul>	ing Data Science; Cathy O'Neil and Rachel Schutt; O'Reilly Media Inc.	

Datasets may be downloaded from the website "http://www1.aucegypt.edu/faculty/hadi/RABE5/"

	CAN-E305A Hours / Week: 3	Paper: Information Retrieval Total Contact Hours: 30	Credit: 3
Course O		Total Contact Hours: 50	Creat: 5
		this course, students will be able to:	
	earn the information		
✓ B	e familiar with Web	Search Engine.	
	e exposed to Link A		
✓ U	nderstand Hadoop a	nd Map Reduce.	
✓ L	earn document text 1	nining techniques.	
UNITS		COURSE CONTENT	
	Introduction		(6L)
1		ory of IR- Components of IR - Issues –Open source Search engine	
		The role of artificial intelligence (AI) in IR – IR Versus Web Sear	ch - Components of a Search
	engine- Character		
			(6L)
2		ctor-space retrieval models- Term weighting - TF-IDF wei Inverted indices - efficient processing with sparse vectors – I	
		Latent Semantic Indexing - Relevance feedback and query expansion	
		ine – Introduction And Crawling	(6L)
		view, web structure, the user, paid placement, search engine of	
3		earch engine optimization/spam – Web Search Architectures	
		g - web indexes — Near-duplicate detection - Index Compression -	
		nk Analysis And Specialized Search	(6L)
		ubs and authorities - Page Rank and HITS algorithms -Searchin	
4	Scoring and rank	ting for Web – Similarity - Hadoop & Map Reduce - Evalua	tion - Personalized search -
		ering and content-based recommendation of documents and pro	
		neration, Summarization, Question Answering, Cross- Lingual Ret	
-	Document Text		(6L)
5		ing; organization and relevance feedback – Text Mining -Text c	
		Igorithms: naive Bayes; decision trees; and nearest neighb	or - Clustering algorithms:
Reference		stering; k-means; expectation maximization (EM).	
		an, and H. Schütze, "Introduction to Information Retrieval", Cambes and Berthier Ribeiro - Neto, "Modern Information Retrieval:	
	behind Search", AC		The Concepts and Technolog
		M Press Books. I Metzler and Trevor Strohman, "Search Engines: Information R	atriaval in Practice" Addiso
	Wesley.	i metzier and rievor Suomnan, Search Englies: Information R	euleval III Flactice, Addiso
	•	ntroduction to Search Engines and Web Navigation", Edition Wile	V
		narles L. A. Clarke, Gordon V. Cormack, "Information Retrieval:	
	Search Engines", Th		implementing and Evaluatin
			· 10 · "00 ·

- •
- Ophir Frieder "Information Retrieval: Algorithms and Heuristics: The Information Retrieval Series", Springer. Manu Konchady, "Building Search Applications: Lucene, Ling Pipe", and First Edition, Gate Mustru Publishing. •

	CAN-E305BPaper: Data Warehousing and Data MiningHours / Week: 3Total Contact Hours: 30	Credit: 3
Course Ou		
After succe	essful completion of this course, students will be able to:	
	tudy of different sequential pattern algorithms	
✓ St	tudy the technique to extract patterns from time series data and it application in real world.	
	an extend the Graph mining algorithms to Web mining	
✓ H	elp in identifying the computing framework for Big Data	
UNITS	COURSE CONTENT	
	Introduction to Data Warehousing:	(6L
1	The need for data warehousing, Operational and informational Data stores, Data wareh	
1	characteristics, Data warehouse architecture, Data warehouse Database, Sourcing, Acqui	
	transformation tools, Metadata, Access tools, Data marts, Data warehousing administration a	nd management.
	Online analytical processing (OLAP):	(4I
2	Need for OLAP, Multidimensional data model, OLAP guidelines, Multidimensional vs. Muilti-relational	
	(OLAP), Categorization of OLAP tools, OLAP tools internet.	
	Introduction to data mining:	(6L
3	The motivation, Learning from past mistake, Data mining, Measuring data mining effective	
5	mining into business process, What is decision tree, Business score card, Where to use decis	ion tree, The general
	idea, How the decision tree works.	
	Classification and prediction:	(5L
4	Cluster Analysis – Types of Data in Cluster Analysis, Partitioning methods, Hierarchical Me	thods; Transactional
	Patterns and other temporal based frequent patterns	
-	Time Series Analysis:	(4L
5	Time series Data, Periodicity Analysis for time related sequence data, Trend analysis, Simila series analysis.	arity search in Time-
	Web Mining:	(5L
6	Web Mining, Mining the web page layout structure, mining web link structure, mining mu	
0	web, Automatic classification of web documents and web usage mining; Distributed Data M	
eference F		

Data Warehousing Fundamentals for 11 Professionals, Second Edition by PaulrajPonnian, Wiley India.
 Principles and Implementation of Data Ware housing, Rajeev Parida Fire Wall Media, Lakshmi Publications.2006.

	CAN-E305C Hours / Week: 3	Paper: Introduction to Big Data Analytics Total Contact Hours: 30	Credit: 3
Course O			
		is course, students will be able to:	
		cisions and create competitive advantage with Big Data	
	nalytics		
		concepts of big data analytics.	
		data using intelligent techniques.	
		earch methods and visualization techniques.	
		iniques for mining data stream.	
		ons using Map Reduce Concepts.	
	troduce programming	tools PIG & HIVE in Hadoop echo system.	
UNITS	<b>T</b> ( <b>1</b> ( <b>1</b> )	COURSE CONTENT	
1	Introduction to big		(4L)
1		Data Platform – Challenges of Conventional Systems - Int	telligent data analysis – Nature of
		esses and Tools - Analysis vs Reporting.	
	Mining data stream		(6L)
2		ams Concepts – Stream Data Model and Architecture - Str ring Streams –Counting Distinct Elements in a Stream –	
Z		w – Decaying Window - Real time Analytics Platform(RTA	
		t Analysis- Stock Market Predictions.	AI ) Applications – Case Studies -
	Hadoop	t Analysis- Stock Market I redictions.	(8L)
		Hadoop Distributed File System, Components of Hadoop	
3		p Streaming- Design of HDFS-Java interfaces to HDFS B	
5		ap Reduce Works-Anatomy of a Map Reduce Job run-Fai	
		n - Map Reduce Types and Formats- Map Reduce Features	
	Frameworks		(6L)
4		Data Using Pig and Hive – Data processing operators in	
		ve - fundamentals of HBase and ZooKeeper - IBM InfoSph	
	Predictive Analytic		(6L)
5	Simple linear regres	sion, Multiple linear regression, Interpretation of regression	
	Visualizations - Vis	al data analysis techniques- interaction techniques - Systen	ns and applications.
Reference l	Books:		
		e Guide, Tom White Third Edition, O'reilly Media, 2012.	
•	Understanding Big Da	ta: Analytics for Enterprise Class Hadoop and Streaming	Data, Chris Eaton, Dirk DeRoos,
		Lapis, Paul Zikopoulos, McGrawHill Publishing, 2012.	010
•	Nining of Massive Dat	asets, Anand Rajaraman and Jeffrey David Ullman, CUP,20	U12.
•	Franks, John Wiley& s	Tidal Wave: Finding Opportunities in Huge Data Stream	is with Advanced Analytics, Bill
	i ranks, joint whey a s		

Making Sense of Data, Glenn J. Myatt, John Wiley & Sons, 2007.

	CAN-E305D Hours / Week: 3	Paper: Graph Theory Total Contact Hours: 30	Credit: 3
Course (	Outcome:		
		this course, students will be able to:	
		arate mathematical definitions of objects in graph theory.	
		nitions to identify and construct examples and to distinguish example	es from non-examples.
		assess a mathematical proof.	,· · ,· ,· ,
		f theoretical knowledge and independent mathematical thinking in	n creative investigation of
	questions in graph the		
	Reason from definition	ns to construct mathematical proofs.	
UNITS	Introduction:	COURSE CONTENT	(6L)
1	Graph, Application Graph, Isomorphi	n of Graph, Finite and Infinite Graph, Incidence & Degree, Isolate sm, Subgraphs, Walks, Paths, and Circuits, Connected Graphs, I er Graphs, Operations On Graphs, Hamiltonian Paths and Circuits	d & Pendant Vertex, Null Disconnected Graphs, and
	Trees		(6L)
2	Spanning Tree, F Separability, Relat	Distance and Centres, Types of Tree, Tree Enumeration, Label undamental Circuits, Cut Sets, Properties, Fundamental Circuit a ed Theorems.Spanning trees, Fundamental circuits, Spanning trees in et, All cut sets, Fundamental circuits and cut sets, Connectivity and se	ed Tree, Unlabeled Tree, and Cut-set, Connectivity, a weighted graph, cut sets,
	Connectivity And		(6L)
3	Network Flows, I	Planar Graph, Representation, Detection, Dual Graph, Geometric, Digraph, Properties, Euler Digraph.	
	Matrices, Colouri	ing	(6L)
4		tion, Adjacency matrix, Incidence matrix, Circuit matrix, Cut-set mat ns – Correlations. Graph Coloring, Chromatic Polynomial, Chroma Theorems	
	Graph Theoretic		(6L)
5	Graph Algorithms	s- Connectedness and Components- Spanning Tree- Fundamenta Shortest Path – Applications overview.	
Reference			
•		n Theory: With Application to Engineering and Computer Science", P	rentice Hall of India.
•		ete and Combinatorial Mathematics: An Applied Introduction", Addis	
•		D.A, "A First Look at Graph Theory", Allied Publishers.	
٠		and Baker T.P. "Discrete Mathematics for Computer Scientists and	Mathematicians", Prentice
	Hall of India.	•	
	Lin CI "Elements	of Disarata Mathematics" Macrow Hill	

- •
- Liu C.L., "Elements of Discrete Mathematics", McGraw Hill. Rosen K.H., "Discrete Mathematics and Its Applications", McGraw Hill. •

	CAN-E305E Hours / Week: 3	Paper: Operation Research and Optimization Techniques Total Contact Hours: 30	Credit: 3
Course O	utcome:		
After succ	cessful completion of t	his course, students will be able to:	
✓ :	Describe the way of w	riting mathematical model for real-world optimization problems.	
✓ I	dentify Linear Program	nming Problems and their solution techniques	
$\checkmark$	Categorize Transporta	tion and Assignment problems	
✓ .	Apply the way in whic	h Game Theoretic Models can be useful to a variety of real-world scenari	os in economics and
i	n other areas.		
√ (	Convert practical situat	ions into non-linear programming problems.	
✓ S	Solve unconstrained an	d constrained programming problems using analytical techniques.	
UNITS		COURSE CONTENT	
	Linear Programm	ing Problem (LPP)-I	(8L)
1	Formulation of an LPP; Graphical Method of solution of an LPP; Convex Combination and Convex Set; Convex		
1	Hull and Convex	Polyhedron; Canonical and Standard form of an LPP; Basic Solution	of a system of linea
	equations; Simplex	Method; Big-M Method; Concept of Duality; Mathematical formulation of	of duals.
		ing Problem (LPP)-II	(8L)
2		blems (TP); Representation of Transportation Problems as LPP; Method	
2		f TP: North-West Corner Rule, Matrix Minima Method, Vogel's A	Approximation Method
	Optimality test of the	ne basic feasible solution; Assignment Problems; Hungarian Method.	
	Game Theory		(7L)
3		gies; The Minimax and Maximin Criterion; Existence of Saddle Point;	
5		h saddle Point - Pure Strategies; Games without a Saddle Point - Mixed	d Strategies; Symmetri
		Principle; Graphical Method of Solution; Algebraic Method of Solution.	
		amming Problem (NLPP)	(7L)
4		timization; Multivariate Optimization with no constraints: Semidefini	
•		ization with Equality Constraints: Method of Lagrange Multipliers; Mu	ltivariable Optimizatio
		straints: Kuhn-Tucker Conditions.	
Reference			
•		and Game Theory by J. G. Chakraborty and P. R. Ghosh, Moulik Library.	
•		by KantiSwarup, P. K. Gupta and Man Mohan, S. Chand and Sons.	
•		tion by S. S. Rao, New Age Techno Press.	
•	Operations Research	by J K Sharma, Macmillan India Ltd	

Master	of Computer Appli	cation	
	CAN-E305F Hours / Week: 3	Paper: Pattern Recognition Total Contact Hours: 30	Credit: 3
$\begin{array}{c} \checkmark & 1 \\ \checkmark & 1 \end{array}$	cessful completion of t dentify where, when a Equipped with basic m Understand a variety o Apply machine learnin Design and develop a p	this course, students will be able to: and how pattern recognition can be applied. athematical and statistical techniques commonly used in pattern rec f pattern recognition algorithms. g concepts in real life problems. pattern recognition system for the specific application ution of the pattern recognition system.	ognition
UNITS		COURSE CONTENT	
1	Probability: independent of the second secon	<b>lity, Random Processes and Linear Algebra</b> endence of events, conditional and joint probability, Bayes theorem processes, Expectation, Autocorrelation, Cross-Correlation, spectra.	
2	Linear Algebra Inner product, oute	r product, inverses, eigen values, eigen vectors, singular values, sing	(2L) gular vectors.
3	Bayes Decision Theory       (4L)         Minimum-error-rate classification. Classifiers, Discriminant functions, Decision surfaces. Normal density and discriminant functions. Discrete features.		
4	case. Unsupervised Hierarchical and ot parameter estimati	ood estimation :Gaussian case. Maximum a Posteriori estimation. I l learning and clustering - Criterion functions for clustering. Algori her methods. Cluster validation. Gaussian mixture models, Expecta on. Maximum entropy estimation. Sequential Pattern Recogniti HMMs. Continuous HMMs. Nonparametric techniques for d	thms for clustering: K-Means, tion-Maximization method for ion. Hidden Markov Models
5	analysis. Eigen ve	<b>duction</b> nt analysis - it relationship to eigen analysis. Fisher discriminant ctors/Singular vectors as dictionaries. Factor Analysis, Total va Non negative matrix factorisation - a dictionary learning method.	
6	Linear discrimina Gradient descent pr	nt functions rocedures, Perceptron, Support vector machines - a brief introduction	n. (2L)
7		etworks ron - feedforward neural network. A brief introduction to deep n current neural networks.	(4L) eural networks, convolutional
8	Non-numeric data Neighbour method	ods for pattern classification or nominal data. Decision trees: Classification and Regressio	n Trees (CART). K-Nearest
Reference			
•	Christopher M. Bisho Springer, January 200 T. Hastie , R. Tibshir	er E. Hart, David G. Stork, "Pattern Classification", 2/E, Wiley - Intop :, "Pattern Recognition And Machine Learning (Information Scient) ani, J. H. Friedman:, "The Elements of Statistical Learning", 1/E, Sp pp ; "Pattern Recognition and Machine Learning", Springer, 2006	nce and Statistics)",1/E,
•	Shigeo Abe, "Advance	ces in Pattern Recognition", Springer, 2005 downloaded from the website "http://www1.aucegypt.edu/faculty/l	hadi/RABE5/"

Course Outcome:         After successful completion of this course, students will be able to:         ✓       Understand the concept of machine learning.         ✓       Identify the regression and classification problem.         ✓       Relate the supervised, unsupervised learning in the real life problem.         ✓       Evaluate the machine learning models with respect to the performance parameters.         ✓       Design and implement various machine learning algorithms in the range of real world problems.         UNITS       COURSE CONTENT         1       Introduction to Machine Learning         1       Introduction to Artificial Intelligence, Machine Learning         2       Scalar, Vector, Matrix, Matrix Operation, Norms, Probability, Joint Distribution, Bayes Theorem, Expectation, C         1       Introduction to Classification         3       Simple Linear Regression, Multiple Linear Regression, Least square gradient descent, Linear Classification, Logistic Regression         4       Regression and Classification for simple trees and computational complexity. Overfitting, noisy data, and pruning.         5       Bagging, boosting, and DECORATE. Active learning with ensembles.         6       Neurons and biological motivation. Linear threshold units. Perceptrons: representational limitation and gradient descent training. Multilayer networks and backpropagation. Hidden layers and constructing intermediate, distributed representations. Overfitting, learning network st	Code: MCAN-305G Contacts Hours / Week: 3		Paper: Machine Learning Total Contact Hours: 30 Credit	it: 3
After successful completion of this course, students will be able to: <ul> <li>Understand the concept of machine learning.</li> <li>Identify the regression and classification problem.</li> <li>Relate the supervised, unsupervised learning in the real life problem.</li> <li>Fevaluate the machine learning models with respect to the performance parameters.</li> <li>Design and implement various machine learning algorithms in the range of real world problems.</li> </ul> <li>UNITS         <ul> <li>COURSE CONTENT</li> <li>Introduction to Machine Learning</li> <li>Introduction to Artificial Intelligence, Machine Learning, Deep Learning</li> <li>Types of Machine Learning</li> <li>Interar Algebra</li> <li>Scalar, Vector, Matrix, Matrix Operation, Norms, Probability, Joint Distribution, Bayes Theorem, Expectation, Cayriance.</li> <li>Regression and Classification</li> <li>Guistice Regression, Multiple Linear Regression, Least square gradient descent, Linear Classification, Logistice Regression</li> <li>Decision Tree Learning</li> <li>Ensemble Learning</li> <li>Ensemble Learning</li> <li>Ensemble Learning</li> <li>Mergensenting concepts as decision trees. Recursive induction of decision trees. Picking the best splitting attribute entropy and information gain. Searching for simple trees and computational complexity. Overfitting, noisy data, and pruning.</li> </ul> </li> <li>Ensemble Learning</li> <li>Meany Intera Networks</li> <li>Meany Intera Networks</li> <li>Maximum margin linear separators. Quadratic programming solution to finding maximum margin separators.</li> <li>Kernels for learning model classified data. Clustering. Hierarchical Aglomerative Clustering, k-means partitional clustering Expe</li>				
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9       Bagging, boosting, and DECORATE. Active learning with ensembles.       (4)         6       Artificial Neural Networks       (4)         6       Neurons and biological motivation. Linear threshold units. Perceptrons: representational limitation and gradient descent training. Multilayer networks and backpropagation. Hidden layers and constructing intermediate, distributed representations. Overfitting, learning network structure, recurrent networks.       (4)         7       Support Vector Machines       (4)         8       Support Vector Machines       (4)         8       Probability theory and Bayes rule. Naive Bayes learning algorithm. Parameter smoothing. Generative vs. discriminative training. Logistic regression. Bayes nets and Markov nets for representing dependencies.       (4)         9       Clustering and Unsupervised Learning       (2)         9       Learning from unclassified data. Clustering. Hierarchical Aglomerative Clustering. k-means partitional clustering. Expectation maximization (EM) for soft clustering. Semi-supervised learning with EM using labeled and unlable data.       (4)         10       Dimensionality Reduction       (4)         10       Principal component Analysis(PCA), Linear Discriminant Analysis(LDA), Feature selection, Feature manipulation and normalization         efference Books:       •       Pattern Recognition and Machine Learning- Christopher M. Bishop, Springer         •       The Elements of Statistical Learning: Data Mining, Inference, and Prediction - Trev				()
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Machine Learning -Tom M. Mitchell, Mc Graw Hill				

	CAN-E394A Paper: Image Processing Lab Hours / Week: 4 Total Contact Hours: 40	Credit: 2
Course O		
After succ	essful completion of this course, students will be able to:	
	tudents will learn to convert one image form to another image form.	
	Able to learn various kinds of image enhancement and image restoration techniques.	
	They will learn various techniques of image compression, image segmentation etc.	
UNIT	COURSE CONTENT	
1	Display of Grayscale Images	
2	Histogram Equalization	
3	Non-linear Filtering	
4	Edge detection using Operators	
5	2-D DFT and DCT	
6	Filtering in frequency domain	
7	Filtering in spatial domain	
8	Display of color images	
9	Discrete Wavelet Transform (DWT) of images	
10	Segmentation using watershed transform	
11	Image Compression	
12	Applications of image zooming and image shrinking etc	

	CAN-E394B Hours / Week: 4	Paper: Web Enabled JAVA Programming LAB Total Contact Hours: 40	Credit: 2
Course O	Outcome:		
After succ	cessful completion of th	nis course, students will be able to:	
✓ (	Create dynamic Websit	e/ Web based Applications	
UNIT		COURSE CONTENT	
1	HTML to Servlet A	pplications	
2	Applet to Servlet Co	ommunication	
3	Designing online applications with JSP		
4	Creating JSP program using JavaBeans		
5	Working with Enter	prise JavaBeans	
6	Performing Java Da	tabase Connectivity.	
7	Creating and Sendir	g Email with Java	
8	Building web applic	ations	

	CAN-E394C Paper: Cloud Computing Lab			
<u>Contacts</u> Course C	Hours / Week: 4 Total Contact Hours: 40	Credit: 2		
	cessful completion of this course, students will be able to:			
	Adapt different types of virtualization and increase resource utilization.			
	Build a private cloud using open source technologies.			
	Analyze security issues on cloud.			
	Develop real world web applications and deploy on commercial cloud.			
	Demonstrate various service models.			
UNIT	COURSE CONTENT			
1	Study of NIST model of cloud computing			
	Understand different types of virtualizations, Host and bare metal hypervisor	rs and implement horizontal		
2	scalability.			
	Technology: XEN/ Vmwares EXSi			
2	Implement IaaS using your resources.			
3	Technology: Open Stack / Eucalyptus			
4	Simulate identity management in private cloud			
4	Technology: Open Stack			
5	Explore Storage as a Service for remote file access using web interface.			
3	Technology: ownCloud			
6	Understand security of web server and data directory			
6	Technology: ownCloud			
7	Deploy Platform as a Service; web applications on commercial cloud.			
/	Technology: Google appEngine/ Windows Azure			
	To create and access VM instances and demonstrate various components	such as EC2, S3, Simple DE		
8	DynamoDB	-		
	Technology: Amazon Web Services			
0	Understand on demand application delivery and Virtual desktop infrastructu	re (Software as a Service)		
9	Technology: Ulteo	`````		
10	Understanding of implementation/applications of basic fog computing.			

Master of Co	mputer Application	
Code: MCAN		C 114. 2
Contacts Hou Course Outc		Credit: 2
<ul> <li>✓ After varial Imple</li> <li>✓ How</li> </ul>	successful completion of this course, students will be able to understand the underlying a bles, constants, operators, expressions, HTML Form creation and submissions. POST ementation of Decision, Loops, Functions, Array and Exception Handling concepts. HTML forms are submitted with PHP Server.	
	y to check validation using JavaScript & JQuery.	
	ecting Forms using AJAX Concept.	
UNITS	COURSE CONTENT	
	Fundamental of Web Design	
1	<b>HTML:</b> Introduction, Editor(VS Code/ Sublime), Element, Attribute, Head, Head Formatting, Quotation, Comment, Color, CSS, Link, Image, Table, List, Block & Inline, C File path, Layout, Code, Entity, Symbol, Emoji, Charset, Forms, Form Attributes, Elem Attributes.	lass, ID, Iframe, Script,
-	<b>CSS:</b> Introduction, Selector, External-Internal-Inline CSS, Comments, Color, Backgradding, Height, Width, Box model, Outline, Text, Font, Icon, Link, List, Table, Display Overflow, Float, Inline-bock, Align, Pseudo-class, Pseudo-element, Opacity, Navigation I gallery, Image sprites, Text Effect, Web Fonts, Transition, Animation, Tooltip, Style Image	y, Max width, Position, Bar, Dropdowns, Image
2	<ul> <li>Advance Web Design</li> <li>CSS Responsive Design: Introduction, Viewport, Grid view, Media queries, Responsivideo.</li> <li>Bootstrap: Introduction, Container, Grid, Typography, Color, tables, Images, jumbotron, group, Badges, Progress bar, Spinner, Pagination, List group, Card, Dropdown, Collapse, Input, Input group, Carousel, Modal, Tooltip, Popover, Toast, Scrollspy, Flex, Media object</li> </ul>	, Alerts, Button, Buttor , Navs, Navbar, Forms
3	Introduction to Web Technology & implementation of PHP Programs &Knowi Strings and Functions.	
	Implementing basic PHP programs with Form, Loop, Functions Array and Strings.	
5	Handling Html Form With PHP: Capturing Form. Data Dealing with Multi-value files. Generating File uploaded form. R submission.	Redirecting a form after
6	Database Connectivity with MySql:Programs implementing displaying data from MYSQL to HTML forms using PHP.Programs implementing updating data from MYSQL to HTML forms using PHP.Programs implementing deleting data from MYSQL to HTML forms using PHP.	
7	Java Script & JQuery: Validating forms using JAVASCRIPT.	
8	<b>Connecting Forms using AJAX Concept:</b> Fetching data from one form to another form using AZAX.	

Master of	f Computer Application		
	CAN-E394EPaper: Android Application Development LabHours / Week: 4Total Contact Hours: 40Credit: 2		
Course O	lutcome:		
After succ	cessful completion of this course, students will be able to:		
✓ I	Learn to use Android Application development platform.		
✓ 1	Fo create simple android application		
✓ 1	Fo understand and implement various designing components of Android user interfaces		
✓ 1	Fo design application's main navigation screen		
✓ 1	Fo understand and designing Android Notification (including push notification)		
✓ 1	Fo connect android application to database for data insertion and retrieval		
UNITS	COURSE CONTENT		
1	Writing First ApplicationCreating Android Project, Android Virtual Device Creation, Set up debugging environment, Workspace set up for development, Launching emulator, debugging on mobile devices.		
2	Basic UI design Basics about Views, Layouts, Resources, Input controls, Input Events, Toasts.		
3	More UI Design Layouts design GridView and ListView, Action bar, Adapters, Menus: Option menu, context menu, sub menu, Pickers - Date and Time, Spinners.		
4	Activity and Fragment Activity, Fragment, Activity Lifecycle and Fragment Lifecycle.		
5	Intents Implicit Intents, Explicit intents, communicating data among Activities.		
6	Navigation Drawer           Panel that displays the app's main navigation screens on the left edge of the screen		
7	Android Notifications Toast, Dialogs (TimePicker, DatePicker, Progress, Alert), Notification Manager and Push Notification		
8	<b>Introducing SQLite</b> SQLiteOpenHelper and creating a database - Opening and closing a database, Working with cursors Inserts, updates, and deletes		

	CAN-E394F Paper: Basic Data Science Lab Hours / Week: 4 Total Contact Hours: 40	Credit: 2		
Course Oi		Credit: 2		
	essful completion of this course, students will be able to:			
	erform the quantitative and qualitative analysis of the data.			
	Lealized the basic trends in two variable plots of numerical data.			
	Compute the mean, median, mode, standard deviation, and variance of grouped data			
	Determine the equation of the trend line to forecast outcomes for time periods in the			
	me periods if necessary.	, 8 8		
	Use a computer to develop a regression analysis, and interpret the output that is asso	ciated with it.		
	Construct machine learning models for providing business ideas.			
UNITS	COURSE CONTENT			
1	Read the datasets(.txt, .xlsx, .csv) from the local system			
2	Make numerical summery(descriptive statistics) of data	Make numerical summery(descriptive statistics) of data		
3	Apply various measures- range, interquartile range, mean absolute deviation, var	Apply various measures- range, interquartile range, mean absolute deviation, variance, and standard deviation		
4	Make graphical summery(histogram, scatterplot, pie plot, boxplot) of data			
5	Detect the outliers(if exists), impute the outliers using suitable methods.			
6	Implement simple linear regression with suitable datasets, observe the standard error, p-value, R-square values.			
	Implement the multivariate regression with suitable datasets and present the outp			
	(a) What percentage of variation in the response is explained by these predictors	?		
7	(b) Which observation has the largest (positive) residual? Give the case number.			
/	(c) Compute the mean and median of the residuals.			
	(d) Compute the correlation of the residuals with the fitted values.			
	(e) Compute the correlation of the residuals with the target variable.			
8	Implement the logistic regression using suitable datasets. Evaluate the performance of the model.			
9	Implement kNN for classification and regression problem.			
	Implement weighted kNN for classification and regression problem.			
<u>10</u> 11	Implement decision tree and regression tree and check the model performance.			

Master of Computer Application	ion	
Code: MCAN-381 Contacts Hours / Week: 8	Paper: Minor Project and Viva-voce Total Contact Hours: 80	Credit: 5
A student needs to pursue	e a research/application based project in his/her institution under ned teacher(s) and on completion of the same an evaluation will be made on	the direct
Project Report, Project Presenta		



# **MASTER OF COMPUTER APPLICATION**

Syllabus w.e.f. the Academic Session 2020-2021





MAULANA ABUL KALAM AZAD UNIVERSITY OF TECHNOLOGY WEST BENGAL

#### Second Year: Semester-IV

Code: MCAN-O401	Paper: Open Elective	
Contacts Hours / Week: Not applicable	Total Contact Hours: Not applicable	Credit: 3
Open Elective		
Business Analytics		
Robotics		
Bioinformatics		
Information Theory &Coding		
Automation in VLSI Design		
Intelligent Control		
Design of Embedded Systems		
Machine Learning		
Soft Computing		
Information Retrieval		
Distributed System		
Digital Marketing		
(While opting for a domain for pursuit	ng the online course, a student needs to ensure that the	e domain was not covered in
previous semesters of the program.)		
Student may opt for a minimum 12-we	ek online course from Massive Open Online Courses (M	MOOCs) in one or any of the
above domains with prior approval for	m the University and/or his/her Institution. In such cas	se, student needs to submit a
self-attested copy of the mark-sheet of	this online course to college well before the end of S	Semester IV. Directly on the

basis of the result obtained by a student, final marks will be allocated to him/her. In case a student pursues a non-

evaluation based online course, an in-house evaluation would be done.

MCA Syllabus

Code: MCAN-481	Paper: Comprehensive Viva-Voce	
Contacts Hours / Week: Not A	Applicable Total Contact Hours: Not Applicable	Credit: 2

#### **Course Objective**

Objective of comprehensive viva-voce is to assess the overall knowledge of the student in the relevant field of computer science and application acquired over 2 years of study in the MCA Program.

The viva shall normally cover the subjects taught in all the semesters of MCA Program. This will test the student's learning and understanding during the course. In doing so, the main objective of this course is to prepare the students to face interview both in the academic and the industrial sector.

Code: MCAN-482	Paper: Major Project & Viva-Voce	
<b>Contacts Hours / Week: 28</b>	Total Contact Hours: 12-15 Weeks	Credit: 20
A student needs to pursue a research/application based project in any of the following modes:		

[A] In his/her institution under the supervision/mentorship of assigned teacher(s) belonging to that institution.

[B] In his/her institution under the joint supervision/mentorship of assigned teacher(s) belonging to that institution and invited external expert(s).

[C] In a research/software/hardware organization under the joint supervision/mentorship of assigned teacher(s) belonging to that institution and external expert(s) belonging to that research/software/hardware organization.

On completion of the same, an evaluation will be made by the institution on the basis of Project Report, Project Presentation, Viva-voce and sufficient measures will be taken by the institution to understand that the project is an outcome based work as a product of student's sole effort.