# SEMESTER V

AMT301	DATA HANDLING AND VISUALIZATION	CATEGORY	L	т	Ρ	CREDIT
		PEC	3	1	0	4

**Preamble:** The syllabus is prepared with the view of facilitating the learner to get an overview of data visualization. This course aims at providing fundamental knowledge in various data visualization techniques using R programming language. It also deals with security aspects involved in data visualization. The learner will be able to understand the process and security aspects involved in data visualization and apply the tools in solving complex problems.

# Prerequisite: Nil

**Course Outcomes:** After the completion of the course the student will be able to:

CO 1	Summarize the key techniques and theory used in visualization (Cognitive Knowledge Level : Understand)
CO 2	Design and use various methodologies present in data visualization. (Cognitive Knowledge Level : Understand)
	Employ various processes and tools for data visualization.
CO 3	(Cognitive Knowledge Level : A <mark>p</mark> ply)
CO 4	Use interactive data visualization to make inferences.(Cognitive Knowledge Level : Apply)
CO 5	Recognize the process involved and security issues present in data visualization. (Cognitive Knowledge Level : Understand)

# Mapping of course outcomes with program outcomes

	<b>PO</b> 1	PO 2	PO 3	PO 4	PO 5	PO6	P07	PO 8	PO 9	PO1 0	PO11	PO1 2
CO1	Ø					2	14	<u>/</u>				$\oslash$
CO2	Ø	$\oslash$	$\oslash$		×			1				$\oslash$
CO3	Ø	$\oslash$	$\oslash$	$\oslash$	$\oslash$							Ø
CO4	Ø	Ø	Ø	Ø	$\oslash$							$\oslash$
CO5	Ø	Ø	Ø	Ø								Ø

Abstrac	Abstract POs defined by National Board of Accreditation						
PO#	Broad PO	PO#	Broad PO				
PO1	Engineering Knowledge	P07	Environment and Sustainability				
PO2	Problem Analysis	PO8	Ethics				
PO3	Design/Development of solutions	PO9	Individual and team work				
PO4	Conduct investigations of complex problems	PO10	Communication				
PO5	Modern tool usage	P011	Project Management and Finance				
PO6	The Engineer and Society	PO12	Lifelong learning				

# Assessment Pattern

	Continuous Asses	ssment Tests	EndSemester
Bloom's Category	Test1 (percentage)	Test2 (percentage)	Examination Marks
Remember	40	40	40
Understand	40	40	40
Apply	20	20	20
Analyze			
Evaluate			
Create	Es	td.	

# Mark Distribution

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3

# **Continuous Internal Evaluation Pattern:**

Attendance	10 marks
Continuous Assessment Tests(Average of Internal Tests 1 & 2)	25 marks
Continuous Assessment Assignment	15 marks

#### **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. First Internal Examination shall be preferably conducted after completing the first half of the syllabus and the Second Internal Examination shall be preferably conducted after completing the remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly covered module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 questions (preferably, 3 each with 7 marks. Out of the 7 questions in Part B, a student should answer any 5.

#### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have a maximum of 2 subdivisions and carries 14 marks.

#### Syllabus

#### Module 1(Introduction to Data Visualization)

Introduction to Visualization – Need and purpose, External representation – Interactivity – Difficulty in Validation, Data Abstraction: Dataset types – Attribute types – Semantics, Task Abstraction – Analyse, Produce, Search, Query, Four levels of validation – Validation approaches – Validation examples. Marks and Channels. Data Visualization tools.

#### Module 2 (Arranging Spatial Data and Networks)

Arrange tables: Categorical regions – Spatial axis orientation – Spatial layout density, Arrange spatial data: Geometry – Scalar fields – Vector fields – Tensor fields. Arrange networks and trees: Connections, Matrix views – Containment, Map color: Color theory, Color maps and other channels.

#### Module 3 (Data Visualization using R)

Installing packages in R, Data types in R, Special values in R, Matrices in R, Data frames in R, Importing data in R, Exporting data in R, Writing a function in R, Writing if else statements in R, Basic loops in R, Nested loops in R, The apply, lapply, sapply, and tapply functions, Using par to beautify a plot in R, Saving plots

#### Module 4 (Basic and Interactive Plots)

Basic and Interactive Plots: scatter plot, interactive scatter plot, bar plot, line plot, interactive Gantt/timeline chart, Merging histograms, interactive bubble plot, waterfall plot, Heat Maps and Dendrograms: simple dendrogram, dendrograms with colors and labels, heat map, heat map with customized colors, three-dimensional heat map and a stereo map, tree map. Maps: regional maps, choropleth maps, contour maps, maps with bubbles, Integrating text with maps, shapefiles, cartograms, Pie Chart and Its Alternatives, Adding the Third Dimension: 3D scatter plot, 3D pie chart, 3D histogram, 3D contour plot.

#### Module 5 (Security Data Visualization)

Port scan visualization - Vulnerability assessment and exploitation - Firewall log visualization - Intrusion detection log visualization - Attacking and defending visualization systems – Creating security visualization system.

#### **Text Books**

- Tamara Munzner, Visualization Analysis and Design, AK Peters Visualization Series, CRC Press, Nov. 2014
- 2. Atmajitsinh Gohil, "R Data Visualization Cookbook", PACKT, 2015.
- 3. Greg Conti, "Security Data Visualization: Graphical Techniques for Network Analysis", NoStarch Press Inc, 2007.

# **Reference Books**

- 1. A Julie Steele and Noah Iliinsky, Designing Data Visualizations: Representing Informational Relationships, O'Relly.
- 2. Andy Kirk, Data Visualization: A Successful Design Process, PAKT.
- 3. Nathan Yau, "Data Points: Visualization that means something", Wiley, 2013.
- 4. Scott Murray, "Interactive data visualization for the web", O"Reilly Media, Inc., 2013.

# **Course Level Assessment Questions**

**CourseOutcome 1 (CO1):** Illustrate the four levels of validation in Data Visualization. **CourseOutcome 2(CO2):** Summarize the different methods to arrange spatial data and networks.

**Course Outcome 3(CO3):** Apply the different R programming concepts usingappropriate data types for applications.

**Course Outcome 4 (CO4):** Model the given dataset: [5, 10, 13, 19, 21, 25, 22, 18, 15, 13, 11, 12, 15, 20, 18, 17, 16, 18, 23, 25] using any 4 interactive plots and appropriate visualizing charts.

**Course Outcome 5 (CO5):** Illustrate how email transmission at different levels of network can be visualized.

# **Model Question Paper OP CODE: Reg No:** PAGES:4 Name: APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY FIFTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR **Course Code: AMT 301 Course Name: Data Handling and Visualization** Max. Marks: 100 **Duration: 3 Hours PART A Answer All Questions. Each Question Carries 3 Marks** Define data visualization? Illustrate the need of data visualization. Why do data semantics and data types matter in data visualization? Describe the HSL system. Discuss the different forms of data types in R. Give one example each. Define the significance of data frames in R.

- Describe scatterplot with an example. 6.
- Illustrate with examples how the waterfall plot constructed in R? 7.
- Compare chloropleth maps and cartograms. 8.
- Outline port scan visualization. 9.
- **10.** Explain vulnerability assessment in visualization?

#### (10x3=30)

#### Part B

#### (Answer any one question from each module. Each question carries 14 Marks)

"Splitting the complex problem of visualization design into four cascading **11.** (a) (10)levels provides an analysis framework that lets you address different

- 1.
- 2.
- 3.
- 4.
- 5.

concerns separately". Explain the four cascading levels with a diagram?

(b)	List the threats to validate each of the levels?	(4)
-----	--	-----

#### OR

12.	(a)	Define marks and channels. Explain how visual channels control the appearance of marks. How are these visual channels and marks used for encoding different chart types?	(8)
	(b)	Illustrate the different data visualization tools.	(6)
13.	(a)	Distinguish between node-link diagrams and matrix views. Also specify the costs and benefits of each.	(8)
	(b)	Explain Treemaps and GrouseFlocks.	(6)
		OR	
14.	(a)	Summarize color mapping. Explain the different types of color maps.	(8)
	(b)	Explain scalar fields, vector fields and tensor fields.	(4)
15.	(a)	Implement a R program to count the number of NA values in a data frame column.	(7)
	(b)	Implement a R program to concatenate two given matrices of same column but different rows.	(7)
		OR	
16.	(a)	Explain in detail how the par() function can be used to beautify plots in R.	(7)
	(b)	Implement a R program to extract every nth element from a vector.	(7)
17.	(a)	Define is a dendrogram? Write the R code to construct a dendrogram.	(7)
	(b)	Explain pie chart, its limitation and the steps involved for its construction in R.	(7)
		OR	
18.	(a)	Illustrate the need of a 3D scatter plot? Write the sample code to generate a	(7)

(b) Illustrate how and where are shape files used and the step-by-step procedure (7) to construct a shape file.

3D scatter plot in R.

(7)

<b>19.</b> (a)	Summarize firewall log visualization.	(7)
----------------	---------------------------------------	-----

(b) Discuss in detail about intrusion detection log visualization. (7)

# OR

- **20.** (a) Describe the concept of attacking and defending visualization systems. (7)
  - (b) Describe about security visualization system.

	Teaching Plan	
No	Contents	No. of Lecture Hours (45 hrs)
Mod	ule - 1 (Introduction to Data Visualization)	(10 hours)
1.1	Introduction to Visualization – Need and purpose	1 hour
1.2	Data Abstraction: Dataset types	1 hour
1.3	Attribute types	1 hour
1.4	Semantics	1 hour
1.5	Task Abstraction, Actions	1 hour
1.6	Targets, Analysing and Deriving	1 hour
1.7	Four levels of validation	1 hour
1.8	Angles to attack, Threats to Validity	1 hour
1.9	Validation approaches	1 hour
1.10	Data Visualization tools	1 hour
	Module - 2 (Arranging Spatial Data and Networks)	(9 hours)
2.1	Arrange tables: Categorical regions – Spatial axis orientation	1 hour
2.2	Spatial layout density	1 hour
2.3	Arrange spatial data: Geometry – Scalar fields	1 hour
2.4	Vector fields	1 hour
2.5	Tensor Fields	1 hour
2.6	Arrange networks and trees: Connections	1 hour
2.7	Matrix views – Containment	1 hour
2.8	Map color: Color theory	1 hour
2.9	Color maps and other channels	1 hour

	Module- 3 (Data Visualization using R)	(8 hours)
3.1	Installing packages in R	1 hour
3.2	Data types in R, Special values in R	1 hour
3.3	Matrices in R, Data frames in R	1 hour
3.4	Importing data in R, Exporting data in R	1 hour
3.5	Writing a function in R, Writing if else statements in R	1 hour
3.6	Basic loops in R, Nested loops in R	1 hour
3.7	The apply, lapply, sapply, and tapplny fuctions, Using par to beautify a plot in R	1 hour
3.8	Saving plots ,Drawing with data	1 hour
	Module - 4 (Basic and Interactive Plots)	(12 hours)
4.1	Basic and Interactive Plots: scatter plot	1 hour
4.2	Interactive Scatter Plot	1 hour
4.3	Interactive Gantt/timeline chart, Merging histograms	1 hour
4.4	Interactive bubble plot, waterfall plot	1 hour
4.5	Heat Maps and Dendrograms : simple dendrogram, dendrograms with colors and labels	1 hour
4.6	Heat Map, Heat map with customized colors	1 hour
4.7	Three-dimensional heat map and a stereo map, tree map	
4.8	Maps: regional maps, choropleth maps, contour maps	1 hour
4.9	Maps with bubbles, Integrating text with maps, shape files, cartograms	1 hour
4.10	Pie Chart and Its Alternatives	1 hour
4.11	Adding the Third Dimension: 3D scatter plot	1 hour
4.12	3D pie chart, 3D histogram, 3D contour plot	1 hour
	Module- 5 (Security Data Visualization)	(6 hours)
5.1	Port scan visualization	1 hour
5.2	Vulnerability assessment and exploitation	1 hour
5.3	Firewall log visualization	1 hour
5.4	Intrusion detection log visualization	1 hour
5.5	Attacking and defending visualization systems	1 hour
5.6	Creating security visualization system	1 hour

CST	COMPUTER	Category	L	Т	Р	Credit	Year of Introduction
303	NETWORKS	РСС	3	1	0	4	2019

**Preamble:** Study of this course provides the learners a clear understanding of how computer networks from local area networks to the massive and global Internet are built, how they allow computers to share information and communicate with one another. This course covers the physical aspects of computer networks, layers of OSI Reference model, and inter-networking. The course helps the learners to compare and analyze the existing network technologies and choose a suitable network design for a given system.

# Prerequisite: Nil

Course Outcomes: After the completion of the course, the student will be able to

CO#	Course Outcomes
CO1	Explain the features of computer networks, protocols, and network design models ( <b>Cognitive Knowledge: Understand</b> )
CO2	Describe the fundamental characteristics of the physical layer and identify the usage in network communication (Cognitive Knowledge: Apply)
CO3	Explain the design issues of data link layer, link layer protocols, bridges and switches (Cognitive Knowledge: Understand)
CO4	Illustrate wired LAN protocols (IEEE 802.3) and wireless LAN protocols (IEEE 802.11) (Cognitive Knowledge: Understand)
CO5	Select appropriate routing algorithms, congestion control techniques, and Quality of Service requirements for a network ( <b>Cognitive Knowledge: Apply</b> )
CO6	Illustrate the functions and protocols of the network layer, transport layer, and application layer in inter-networking ( <b>Cognitive Knowledge: Understand</b> )

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO1 1	PO12
CO1	Ø	Ø	1	r R	Ų.	U.		A	LF	Ŵ		
CO2				11-	10	)L	0		0	AL		
CO3					V	EF	S		Y			9
<b>CO4</b>	Ø											
CO5												
CO6												

Mapping of course outcomes with program outcomes

Abstract POs defined by Nationa	al Board	l of Accreditation
Broad PO	PO#	Broad PO
Engineering Knowledge	PO7	Environment and Sustainability
Problem Analysis	PO8	Ethics
Design/Development of solutions	PO9	Individual and teamwork
Conduct investigations of complex problems	PO10	Communication
Modern tool usage	<b>PO11</b>	Project Management and Finance
The Engineer and Society 201	PO12	Lifelong learning
	Broad PO       I         Engineering Knowledge       I         Problem Analysis       I         Design/Development of solutions       I         Conduct investigations of complex problems       I         Modern tool usage       I	Engineering KnowledgePO7Problem AnalysisPO8Design/Development of solutionsPO9Conduct investigations of complex problemsPO10Modern tool usagePO11

# **Assessment Pattern**

Bloom's Category	Test 1 (Marks in percentage)	Test 2 (Marks in percentage)	End Semester Examination (Marks in percentage)
Remember	40	30	30

Understand	50	50	50
Apply	10	20	20
Analyze	DDLI	LIZA	A A A
Evaluate	ADDO	LINA	LAM
Create	-NO	OGL	CAL

# **Mark Distribution**

Total Ma	rks	CIE Marks	ESE Marks	ES Dur	E ation
150		50	100	3	

# **Continuous Internal Evaluation Pattern:**

Attendance		: 10 mark <mark>s</mark>
Continuous Asses	sment Test	: 25 mark <mark>s</mark>
Continuous Asses	sment Assignment	: 15 mark <mark>s</mark>

# **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. The first series test shall be preferably conducted after completing the first half of the syllabus. The second series test shall be preferably conducted after completing the remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 questions (preferably, 3 questions each from the 7 questions, a student should answer any 5.

# **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer anyone. Each question can have maximum 2 sub-divisions and carries 14 marks.

# Syllabus

# Module - 1 (Introduction and Physical Layer)

Introduction – Uses of computer networks, Network hardware, Network software. Reference models – The OSI reference model, The TCP/IP reference model, Comparison of OSI and TCP/IP reference models.

Physical Layer – Modes of communication, Physical topologies, Signal encoding, Repeaters and hub, Transmission media overview. Performance indicators – Bandwidth, Throughput, Latency, Queuing time, Bandwidth–Delay product.

# Module - 2 (Data Link Layer)

Data link layer - Data link layer design issues, Error detection and correction, Sliding window protocols, High-Level Data Link Control(HDLC)protocol. Medium Access Control (MAC) sublayer –Channel allocation problem, Multiple access protocols, Ethernet, Wireless LANs - 802.11, Bridges & switches - Bridges from 802.x to 802.y, Repeaters, Hubs, Bridges, Switches, Routers and Gateways.

# Module - 3 (Network Layer)

Network layer design issues. Routing algorithms - The Optimality Principle, Shortest path routing, Flooding, Distance Vector Routing, Link State Routing, Multicast routing, Routing for mobile hosts. Congestion control algorithms. Quality of Service (QoS) - requirements, Techniques for achieving good QoS.

# Module - 4 (Network Layer in the Internet)

IP protocol, IP addresses, Internet Control Message Protocol (ICMP), Address Resolution Protocol (ARP), Reverse Address Resolution Protocol (RARP), Bootstrap Protocol (BOOTP), Dynamic Host Configuration Protocol (DHCP). Open Shortest Path First(OSPF) Protocol, Border Gateway Protocol (BGP), Internet multicasting, IPv6, ICMPv6.

# Module – 5 (Transport Layer and Application Layer)

Transport service – Services provided to the upper layers, Transport service primitives. User Datagram Protocol (UDP). Transmission Control Protocol (TCP) – Overview of TCP, TCP segment header, Connection establishment &release, Connection management modeling, TCP retransmission policy, TCP congestion control.

Application Layer –File Transfer Protocol (FTP), Domain Name System (DNS), Electronic mail, Multipurpose Internet Mail Extension (MIME), Simple Network Management Protocol

(SNMP), World Wide Web(WWW) – Architectural overview.

# **Text Books**

- 1. Andrew S. Tanenbaum, Computer Networks, 4/e, PHI (Prentice Hall India).
- 2. Behrouz A Forouzan, Data Communication and Networking, 4/e, Tata McGraw Hill

# **Reference Books**

- 1. Larry L Peterson and Bruce S Dave, Computer Networks A Systems Approach, 5/e, Morgan Kaufmann.
- 2. Fred Halsall, Computer Networking and the Internet, 5/e.
- 3. James F. Kurose, Keith W. Ross, Computer Networking: A Top-Down Approach, 6/e.
- 4. Keshav, An Engineering Approach to Computer Networks, Addison Wesley, 1998.
- 5. W. Richard Stevens. TCP/IP Illustrated Volume 1, Addison-Wesley, 2005.
- 6. William Stallings, Computer Networking with Internet Protocols, Prentice-Hall, 2004.
- 7. Request for Comments (RFC) Pages IETF -https://www.ietf.org/rfc.html

# **Course Level Assessment Questions**

# **Course Outcome1 (CO1)**

- 1. Compare TCP/IP and OSI reference model.
- 2. The purpose of physical layer is to transport a raw bit stream from one machine to another. Justify.

# **Course Outcome2 (CO2)**

- 1. Write the physical and transmission characteristics of Optical Fibre Cable guided transmission media.
- 2. The distance between the sender and receiver systems is about 200 KM. The speed of transmission is 2GB/s. Find out the propagation time?

# **Course Outcome3 (CO3)**

- 1. Ethernet frames must be at least 64 bytes long to ensure that the transmitter is still going in the event of a collision at the far end of the cable. Fast Ethernet has the same 64-byte minimum frame size but can get the bits out ten times faster. How is it possible to maintain the same minimum frame size?
- 2. What do you mean by bit stuffing?

# **Course Outcome4 (CO4)**

- 1. Draw and explain the frame format for Ethernet.
- 2. Give the differences between CSMA/CD and CSMA/CA protocol.

# **Course Outcome5 (CO5)**

1. Consider the given subnet in which distance vector routing is used, and the vectors just come in to router C as follows: from B: (5, 0, 8, 12, 6, 2); from D: (16, 12, 6, 0, 9, 10);

and from E: (7, 6, 3, 9, 0, 4). The measured delays from C to B, D, and E, are 6, 3, and 5, respectively. What is C's new routing table? Give both the outgoing line to use and the expected delay.



2. Illustrate the leaky bucket congestion control technique.

# **Course Outcome 6 (CO6)**

- 1. How do you subnet the Class C IP Address 206.16.2.0 so as to have 30 subnets. What is the subnet mask for the maximum number of hosts? How many hosts can each subnet have?
- 2. Give the architecture of World Wide Web.

#### Model Question Paper

**QP CODE:** 

# Reg No:

Name:

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY FIFTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

# Course Code: CST 303

# **Course Name : Computer Networks**

Max Marks: 100

#### **Duration: 3 Hours**

# PART-A

# (Answer All Questions. Each question carries 3 marks)

1. What does "negotiation" mean when discussing network protocols in a layered architecture? Give an example.

**PAGES:** 

- 2. Define simplex, half-duplex, and full-duplex transmission modes. Give one example for each.
- 3. Data link protocols almost always put the CRC in a trailer rather than in a header. Why?
- 4. An 8-bit byte with binary value 10101111 is to be encoded using an even-parity Hamming code. What is the binary value after encoding?
- 5. Illustrate the Count to Infinity problem in routing.
- 6. Describe two major differences between the warning bit method and the Random Early Detection (RED) method.
- 7. The Protocol field used in the IPv4 header is not present in the fixed IPv6 header. Why?
- 8. How many octets does the smallest possible IPv6 (IP version 6) datagram contain?
- Can Transmission Control Protocol(TCP) be used directly over a network (e. g. an Ethernet) without using IP? Justify your answer.
- 10. When Web pages are sent out, they are prefixed by MIME headers. Why?

(10x3=30)

#### Part B

# (Answer any one question from each module. Each question carries 14 Marks)

11.	(a)	With a neat diagram, explain Open Systems Interconnection (OSI) Reference Model.	(8)
	(b)	Compare Twisted Pair, Coaxial Cable and Optical Fibre guided transmission media	(6)

#### OR

12.	(a) Consider two networks providing reliable connection-oriented service. One	
	of them offers a reliable byte stream and the other offers a reliable message	(8)
	stream. Are they identical? Justify your answer.	

(b) Sketch the waveform in Manchester and Differential Manchester Encoding (6) for the bitstream 11000110010.

- 13. (a) A bit stream 10011101 is transmitted using the standard CRC method. The generator polynomial is  $\square^3 + I$ . Show the actual bit string transmitted. Suppose the third bit from the left is inverted during transmission. Show that this error is detected at the receiver's end.
  - (b) Explain the working of High-Level Data Link Control (HDLC) protocol.

(6)

(8)

#### OR

14.	(a)	Explain the working of IEEE 802.11 MAC sublayer.	(10)
	(b)	Distinguish between Bridges and Switches.	(4)
15.	(a)	Illustrate Distance Vector Routing algorithm with an example.	(8)

(b) Explain the characteristics of Routing Information Protocol (RIP). (6)

#### OR

16.	(a)	A computer on a 6-Mbps network is regulated by a token bucket. The token bucket is filled at a rate of 1 Mbps. It is initially filled to capacity with 8 megabits. How long can the computer transmit at the full 6 Mbps?	(8)
	(b)	Explain how routing is performed for mobile hosts.	(6)
17.	(a)	Explain the address resolution problem using Address Resolution Protocol (ARP) and Reverse Address Resolution Protocol (RARP)with an example network.	(10)
	(b)	A network on the Internet has a subnet mask of 255.255.240.0. What is the maximum number of hosts it can handle?	(4)
		OR	
18.	(a)	How do you subnet the Class C IP address 195.1.1.0 so as to have 10 subnets with a maximum of 12 hosts in each subnet.	(6)
	(b)	Draw IPv6 Datagram format and explain its features.	(8)
19.	(a)	Distinguish the header formats of Transmission Control protocol (TCP) and User Datagram Protocol (UDP).	(8)

(b) Explain the principal Domain Name System (DNS) resource record types for (6)

IPv4.

# OR

- 20. (a) What is the role of Simple Mail Transfer Protocol (SMTP) in E- mail? (6)
  - (b) With the help of a basic model, explain the working of World Wide Web (8) (WWW).

# **Teaching Plan**

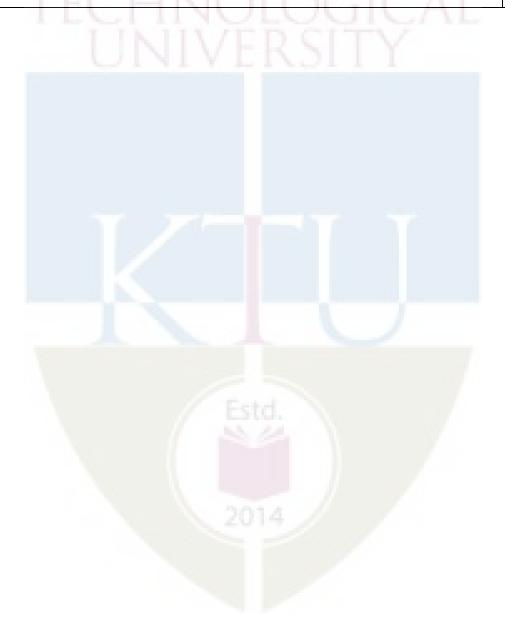
No	Contents	No of Lecture Hrs
	Module – 1 (Introduction and Physical Layer) (10 hrs)	
1.1	Introduction, Uses of computer networks.	1 hour
1.2	Network Hardware, Local Area Networks (LAN), Metropolitan Area Networks (MAN), Wide Area Networks (WAN), Wireless networks, Home networks, Internetworks.	1 hour
1.3	Network Software, Protocol hierarchies, Design issues for the layers.	1 hour
1.4	Connection-oriented and Connectionless services, Service primitives, Relationship of services to protocols.	1 hour
1.5	Reference models, The OSI reference model.	1 hour
1.6	The TCP/IP reference model, Comparisonof OSI and TCP/IP reference models.	1 hour
1.7	Physical layer, Modes of communication, Simplex, Half-duplex, and Full- duplex, Physical topologies, Mesh, Star, Bus, Ring, Hybrid.	1 hour
1.8	Signal encoding, Manchester, Differential Manchester.	1 hour
1.9	Transmission media overview, Guided media (twisted pair, coaxial and fiber optic media), Unguided/wireless media (radio, microwave, and infrared).	1 hour
1.10	Performance indicators, Bandwidth (in Hertz and in Bits per Seconds),	1 hour

	Throughput, Latency (Delay), Queuing time, Bandwidth-Delay product.	
	Module 2 – (Data Link Layer) (10 hrs)	
2.1	Data link layer design issues.	1 hour
2.2	Error detection and correction, Error correcting codes	1 hour
2.3	Error detecting codes.	1 hour
2.4	Sliding window protocols.	1 hour
2.5	High-Level Data Link Control(HDLC) protocol.	1 hour
2.6	Medium Access Control (MAC) sublayer, Channel allocation problem, Multiple access protocols.	1 hour
2.7	Ethernet, Ethernet cabling, Manchester encoding, Ethernet MAC sublayer protocol, Binary Exponential Backoff algorithm.	1 hour
2.8	Ethernet performance, Switched Ethernet, Fast Ethernet, Gigabit Ethernet, IEEE 802.2: Logical Link Control.	1 hour
2.9	Wireless LANs, 802.11 protocol stack, Physical layer, MAC Sublayer protocol, Frame structure.	1 hour
2.10	Bridges &switches, Bridges from 802.x to 802.y, Repeaters, Hubs, Bridges, Switches, Routers, and Gateways.	1 hour
	Module 3 - (Network Layer) (8 hrs)	1
3.1	Network layer design issues.	1 hour
3.2	Routing algorithms, The Optimality Principle, Shortest path routing, Flooding.	1 hour
3.3	Distance Vector Routing.	1 hour
3.4	Link State Routing.	1 hour
3.5	Multicast routing, Routing for mobile hosts.	1 hour

3.6	General principles of congestion control, Congestion prevention policies, Congestion control in virtual circuit subnets.	1 hour
3.7	Congestion control algorithms, Congestion control in Datagram subnets, Load shedding, Jitter control.	1 hour
3.8	Quality of Service, Requirements, Techniques for achieving good Quality of Service.	1 hour
	Module 4 – (Network Layer in the Internet) (9 hrs)	
4.1	Network layer in the Internet, Internet Protocol (IP).	1 hour
4.2	IP Addresses, Subnets, Classless Inter-Domain Routing (CIDR).	1 hour
4.3	IP Addresses, Network Address Translation (NAT).	1 hour
4.4	Internet Control Message Protocol (ICMP), Address Resolution Protocol (ARP), Reverse Address Resolution Protocol (RARP).	1 hour
4.5	Bootstrap Protocol (BOOTP), Dynamic Host Configuration Protocol (DHCP).	1 hour
4.6	Open Shortest Path First (OSPF) protocol.	1 hour
4.7	Border Gateway Protocol (BGP).	1 hour
4.8	Internet multicasting.	1 hour
4.9	IPv6, Header format, Extension headers, Internet Control Message Protocol version 6 (ICMPv6).	1 hour
	Module 5 - (Transport Layer and Application Layer) (8 hrs)	
5.1	Transport Service, Services provided to the upper layers, Transport service primitives. User Datagram Protocol (UDP).	1 hour
5.2	Transmission Control Protocol (TCP), TCP segment header, Connection establishment & release, Connection management modeling.	1 hour
5.3	TCP retransmission policy, TCP congestion control.	1 hour
5.4	Application layer, File Transfer Protocol (FTP).	1 hour

# B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

5.5	Domain Name System (DNS).	1 hour
5.6	Electronic Mail, Multipurpose Internet Mail Extension (MIME).	1 hour
5.7	Simple Network Management Protocol (SNMP).	1 hour
5.8	World Wide Web, Architectural overview.	1 hour



AMT 305	INTRODUCTION TO MACHINE LEARNING	Category	L	Т	Р	Credit
		PCC	3	1	0	4

#### **Preamble**:

This course enables the learners to understand the advanced concepts and algorithms in machine learning. The course covers the standard and most popular supervised learning algorithms such as linear regression, logistic regression, decision trees, Bayesian learning and the Naive Bayes algorithm, basic clustering algorithms and classifier performance measures. This course helps the students to provide machine learning based solutions to real world problems.

**Prerequisite:** Basic understanding of probability theory and linear algebra

# Mapping of course outcomes with program outcomes

CO1	Illustrate Machine Learning concepts and basics of supervised learning concepts. (Cognitive Knowledge Level: <b>Apply</b> )
CO2	Describe dimensionality reduction techniques and supervised learning concepts (regression, linear classification). (Cognitive Knowledge Level: <b>Apply</b> )
CO3	Solve real life problems using appropriate machine learning models and evaluate the performance measures and Illustrate the concepts of Multilayer neural network . (Cognitive Knowledge Level: <b>Apply</b> )
CO4	Illustrate basics of parameter estimation models and the working of classifier SVM classifier model (Cognitive Knowledge Level: Apply)
CO5	Describe unsupervised learning concepts (Cognitive Knowledge Level: Apply)

# Mapping of course outcomes with program outcomes

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	<b>PO10</b>	PO11	PO12
CO1	$\bigcirc$	$\bigcirc$	0	$\bigcirc$								$\bigcirc$
CO2	$\bigcirc$	$\bigcirc$	0	Ø	0	20	4	/				
CO3	$\bigotimes$	$\bigotimes$	$\bigcirc$	0			/					Ø
CO4	$\bigotimes$	$\bigotimes$	$\bigcirc$	$\bigotimes$								Ø
CO5		$\bigcirc$		$\bigcirc$								

Abstract POs defined by National Board of Accreditation							
PO#	Broad PO	PO#	Broad PO				

PO1	Engineering Knowledge	<b>PO7</b>	Environment and Sustainability
PO2	Problem Analysis	PO8	Ethics
PO3	Design/Development of solutions	PO9	Individual and team work
PO4	Conduct investigations of complex problems	PO10	Communication
PO5	Modern tool usage	PO11	Project Management and Finance
PO6	The Engineer and Society	PO12	Life long learning

# Assessment Pattern

Bloom's Category	Continuous A	Assessment Tests	End Semester Examination Marks (%)	
Category	Test 1 (%)	Test 2 (%)		
Remember	30	30	30	
Understand	30	30	30	
Apply	40	40	40	
Analyze				
Evaluate		_		
Create				

# **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3

# **Continuous Internal Evaluation Pattern:**

Attendance	10 marks
Continuous Assessment Tests(Average of Internal Tests 1 & 2)	25 marks
Continuous Assessment Assignment	15 marks

# **Internal Examination Pattern**

Each of the two internal examinations has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed module), as student should answer any 5.

# **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 full questions from each module of which student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.

# **Syllabus**

# Module-1 (Overview of machine learning)

Introduction to Machine Learning, Machine learning paradigms-supervised, semi-supervised, unsupervised, reinforcement learning.

Supervised learning- Input representation, Hypothesis class, Version space, Vapnik-Chervonenkis (VC) Dimension, Probably Approximately Correct Learning (PAC), Noise, Learning Multiple classes, Model Selection and Generalization

# Module-2 (Supervised Learning)

Dimensionality reduction – Subset selection, Principal Component Analysis.

Regression - Linear regression with one variable, Linear regression with multiple variables, solution using gradient descent algorithm and matrix method, basic idea of overfitting in regression. Linear Methods for Classification- Logistic regression, Naive Bayes, Decision tree algorithm ID3.

Case Study: Develop a classifier for face detection.

# Module-3 (Classification Assessment and Neural Networks (NN))

Classification Performance measures - Precision, Recall, Accuracy, F-Measure, Receiver Operating Characteristic Curve(ROC), Area Under Curve AUC. Bootstrapping, Cross Validation.

Perceptron, Neural Network - Multilayer feed forward network, Activation functions (Sigmoid, ReLU, Tanh), Backpropagation algorithm.

# Module-4 (Parameter estimation & SVM Classifier)

Basics of parameter estimation - Maximum Likelihood Estimation(MLE) and Maximum a Posteriori estimation(MAP). Bias-Variance decomposition.

Support Vector Machines - Introduction, Maximum Margin hyperplanes, Mathematics behind Maximum Margin Classification, soft margin SVM classifier, non-linear SVM, Kernels for learning non-linear functions, polynomial kernel, Radial Basis Function(RBF), Kernel Trick.

# Module-5 (Unsupervised Learning)

Ensemble methods, Voting, Bagging, Boosting.

Unsupervised Learning - Clustering Methods -Similarity measures, K-means clustering, Expectation-Maximization for soft clustering, Hierarchical Clustering Methods , Density based clustering.

# **Text Book**

- 1. Ethem Alpaydin, Introduction to Machine Learning, 2nd edition, MIT Press 2010.
- 2. Tom Mitchell, Machine Learning, McGraw-Hill, 1997.

# **Reference Books**

- 1. Christopher Bishop. Neural Networks for Pattern Recognition,Oxford University Press, 1995.
- 2. Kevin P. Murphy. Machine Learning: A Probabilistic Perspective, MIT Press 2012.
- 3. Trevor Hastie, Robert Tibshirani, Jerome Friedman, The Elements Of Statistical Learning, Second edition Springer 2007.
- 4. P. Langley, Elements of Machine Learning, Morgan Kaufmann, 1995.
- 5. Richert and Coelho, Building Machine Learning Systems with Python.
- 6. Mohammed J. Zaki and Wagner Meira, Data Mining and Analysis: Fundamental Concepts and Algorithms, Cambridge University Press, First South Asia edition, 2016.
- 7. Jake VanderPlas, Python Data Science Handbook, O'Reilly Media, 2016
- 8. Davy Cielen, Arno DB Meysman and Mohamed Ali.Introducing Data Science: Big Data, Machine Learning, and More, Using Python Tools, Dreamtech Press 2016.

# **Course Level Assessment Questions**

#### **Course Outcome1 (CO1):**

- 1. Compare different machine learning paradigms with suitable examples.
- 2. Explain (a) Hypothesis space (b) Version space (c) Most General hypothesis

(d) Most specific hypothesis in the context of a classification problem.

- 3. Define VC dimension. Show that an axis aligned rectangle can shatter 4 points in 2 dimensions.
- 4. Explain the concept of PAC learning . Derive an expression for PAC learning in such a way that the selected function will have low generalized error.

5. Distinguish between overfitting and underfitting. How it can affect model generalization?

#### **Course Outcome 2(CO2):**

- Suppose that you are asked to perform linear regression to learn the function that outputs y, given the D-dimensional input x. You are given N independent data points, and that all the D attributes are linearly independent. Assuming that D is around 100, would you prefer the closed form solution or gradient descent to estimate the regressor?
- 2. Suppose you have a three class problem where class label  $y \in 0, 1, 2$  and each training example X has 3 binary attributes  $X_1, X_2, X_3 \in 0, 1$ . How many parameters (probability distribution) do you need to know to classify an example using the Naive Bayes classifier?
- 3. Is principal component analysis a supervised learning problem? Justify your answer
- 4. Explain feature selection and feature extraction method for dimensionality reduction.

Age	Competition	Туре	Class (profit)
Old	Yes	Software	Down
Old	No	Software	Down
Old	No	Hardware	Down
Mid	Yes	Software	Down
Mid	Yes	Hardware	Down
Mid	No	Hardware	Up
Mid	No	Software	Up
New	Yes	Software	Up
New	No	Hardware	Up
New	No	Software	Up

5. Use the ID3 algorithm to construct a decision tree for the data in the following table.

# Course Outcome 3(CO3):

- Classifier A attains 100% accuracy on the training set and 70% accuracy on the test set. Classifier B attains 70% accuracy on the training set and 75% accuracy on the test set. Which one is a better classifier? Justify your answer.
- 2. What are ROC space and ROC curve in machine learning? In ROC space, which points correspond to perfect prediction, always positive prediction and always negative prediction? Why?

- 3. Suppose there are three classifiers A,B and C. The (FPR, TPR) measures of the three classifiers are as follows A (0, 1), B (1, 1), C (1,0.5). Which can be considered as a perfect classifier? Justify your answer.
- 4. Briefly explain Perceptron Network.
- 5. Briefly explain BackPropagation Network.
- 6. Briefly explain one way in which using tanh instead of logistic activations makes optimization easier.
- 7. ReLU activation functions are most used in neural networks instead of the tanh activation function. Draw both activation functions and give a) an advantage of the ReLU function compared to the tanh function. b) a disadvantage of the ReLU function compared to the tanh function.

#### Course Outcome 4(CO4): .

- 1. What are support vectors and list any three properties of the support vector classifier solution?
- 2. Why do you use kernels to model a projection from attributes into a feature space, instead of simply projecting the dataset directly?
- 3. Describe how Support Vector Machines can be extended to make use of kernels. Illustrate with reference to the Gaussian kernel  $K(x, y) = e^{-z}$ , where  $z = (x-y)^p$ .
- 4. A coin is tossed 100 times and lands heads 62 times. What is the maximum likelihood estimate for  $\theta$ , the probability of heads.
- 5. Suppose data  $x_1, ..., x_n$  are independent and identically distributed drawn from an exponential distribution  $exp(\lambda)$ . Find the maximum likelihood for  $\lambda$ .
- 6. Suppose  $x_1, ..., x_n$  are independent and identically distributed(iid) samples from a distribution with density

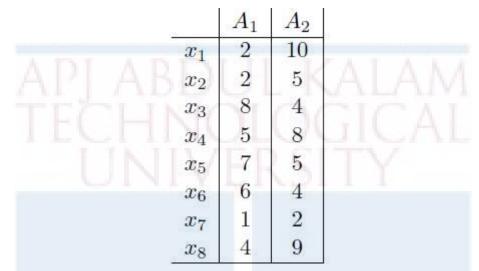
$$f_X(x \mid \theta) = \begin{cases} \frac{\theta x^{\theta - 1}}{3^{\theta}}, & 0 \le x \le 3\\ 0, & \text{otherwise} \end{cases}$$

Find the maximum likelihood estimate(MLE) for  $\theta$ .

7. Find the maximum likelihood estimator (MLE) and maximum a posteriori (MAP) estimator for the mean of a univariate normal distribution. Assume that we have N samples,  $x_1, ..., x_N$ independently drawn from a normal distribution with known variance  $\sigma^2$  and unknown mean  $\mu$  and the prior distribution for the mean is itself a normal distribution with mean vand variance  $\beta^2$ . What happens to the MLE and MAP estimators as the number of samples goes to infinity.

#### Course Outcome 5(CO5): .

- 1. Illustrate the strength and weakness of the K-means algorithm.
- 2. Suppose you want to cluster the eight points shown below using k-means



Assume that  $\mathbf{k} = \mathbf{3}$  and that initially the points are assigned to clusters as follows:

 $C_1 = \{x_1, x_2, x_3\}, C_2 = \{x_4, x_5, x_6\}, C_3 = \{x_7, x_8\}$ . Apply the k-means algorithm until convergence, using the Manhattan distance.

3. Cluster the following eight points representing locations into three clusters:  $A_1(2, 10)$ ,  $A_2(2, 5)$ ,  $A_3(8, 4)$ ,  $A_4(5, 8)$ ,  $A_5(7, 5)$ ,  $A_6(6, 4)$ ,  $A_7(1, 2)$ ,  $A_8(4, 9)$ .

Initial cluster centers are:  $A_1(2, 10)$ ,  $A_4(5, 8)$  and  $A_7(1, 2)$ .

The distance function between two points  $a = (x_1, y_1)$  and  $b = (x_2, y_2)$  is defined as  $D(a, b) = |x_2 - x_1| + |y_2 - y_1|$ 

Use k-Means Algorithm to find the three cluster centers after the second iteration.

- 4. What is ensemble learning? Can ensemble learning using linear classifiers learn classification of linearly non-separable sets?
- 5. Describe boosting. What is the relation between boosting and ensemble learning?

# **Model Question Paper**

# **QP CODE:**

Reg No: \_\_\_\_\_

# Name: \_\_\_\_\_

PAGES:4

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

# FIFTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

# Course Code: AMT 305

# **Course Name: Introduction to Machine Learning**

Max. Marks : 100

**Duration: 3 Hours** 

# PART A

# Answer All Questions. Each Question Carries 3 Marks

- **1.** Distinguish between classification and regression with an example.
- 2. Determine the hypothesis space H and version space with respect to the following data D.

х	2	11	17	0	1	5	7	13	20
Class	0	1	1	0	0	0	0	1	1

- **3.** Is principal component analysis a supervised learning problem? Justify your answer.
- 4. Specify the basic principle of gradient descent algorithm.
- 5. (a)Classifier A attains 100% accuracy on the training set and 70% accuracy on the test set. Classifier B attains 70% accuracy on the training set and 75% accuracy on the test set. Which one is a better classifier? Justify your answer.

(b) How does bias and variance trade-off affect machine learning algorithms?

6. Mention the primary motivation for using the kernel trick in machine learning algorithms?

- Suppose that you have a linear support vector machine(SVM) binary classifier. 7. Consider a point that is currently classified correctly, and is far away from the decision boundary. If you remove the point from the training set, and re-train the classifier, will the decision boundary change or stay the same? Justify your answer.
- Differentiate between bagging and boosting. 8.
- Illustrate the strength and weakness of the k-means algorithm. 9.
- 10. Expectation maximization (EM) is designed to find a maximum likelihood setting of the parameters of a model when some of the data is missing. Does the algorithm converge? If so, do you obtain a locally or globally optimal set of parameters?

(10x3=30)

#### Part B

#### (Answer any one question from each module. Each question carries 14 Marks)

11. (a) Define machine learning. Explain different paradigms of machine learning with (7) examples.

(b) Calculate the VC dimension of the following

1)An open internal in R is defined  $as(a,b) = \{x \in \mathbb{R} \mid a < x < b\}$ . It has two parameters a and b. Calculate the VC dimension of the set of all open intervals.

2) Suppose the instance space X is the set of real numbers and the hypothesis space H is the set of intervals on the real number line. Here, it is evident that H is the set of hypotheses of the form a < x < b, where a and b may be any real constants. What is VC(H)?

#### OR

- Let  $X = R^2$  and C be the set of all possible rectangles in two dimensional plane **12.** (a) (7) which are axis aligned (not rotated). Show that this concept class is PAC learnable.
  - (b) What is meant by noise in data? What are the interpretations of noise? (7)

13. (a) Consider the hypothesis for the linear regression  $h_{\theta}(x) = \theta_0 + \theta_1 x$ , and the cost (7) function J( $\theta_0$ ,  $\theta_1$ ) = 1/2m  $\Sigma_1^{m}(h_{\theta}(x^{(i)}) - y^{(i)})^2$  where m is the number of training examples. Given the following set of training examples.

x	У
3	2
4	2
)	1
ţ.	3

Answer the following questions :

- 1) Find the value of  $h_{\theta}$  (2) if  $\theta_0 = 0$  and  $\theta_1 = 1.5$
- 2) Find the value of J(0,1)
- 3) Suppose the value of J( $\theta_0$ ,  $\theta_1$ ) = 0. What can be inferred from this.
- (b) Let X = R<sup>2</sup> and C be the set of all possible rectangles in two dimensional plane (7) which are axis aligned (not rotated). Show that this concept class is PAC learnable.

#### OR

14. (a) The following dataset can be used to train a classifier that determines whether a given person is likely to own a car or not. There are three features: education level (primary, secondary, or university); residence (city or country); gender (female, male). (7)

education	residence	gender	has car?
sec	country	female	yes
univ	country	female	yes
prim	city	male	no
univ	city	male	no
sec	city	female	no
sec	country	male	yes
prim	country	female	yes
univ	country	male	yes
sec	city	male	yes
prim	city	female	no
univ	city	female	no
prim	country	male	yes

Use ID3 Algorithm and find the best attribute at the root level of the tree

(b) Consider a linear regression problem  $y = w_1x + w_0$ , with a training set having m (7) examples (x1, y1), . . .(xm, ym). Suppose that we wish to minimize the mean 5th degree error (loss function) given by  $1/m \Sigma 1m(y_1 - w_1x_1 - w_0)5$ .

1. Calculate the gradient with respect to the parameter w1.

2. Write down pseudo-code for on-line gradient descent on w1.

3. Give one reason in favor of on-line gradient descent compared to batch-gradient descent, and one reason in favor of batch over on-line.

15. (a) Suppose the dataset had 9700 cancer-free images from 10000 images from cancer (7) patients. Find precision, recall and accuracy ? Is it a good classifier? Justify.

Actual Class\Predicted class	cancer = yes	cancer = no	Total
cancer = yes	90	210	300
cancer = no	140	9560	9700
Total	230	9770	10000

(b) Compare ReLU with Sigmoid function. Consider a neuron with four inputs, and weight of edge connecting the inputs are 1, 2, 3 and 4. Let the bias of the node is zero and inputs are 2, 3, 1, 4. If the activation function is linear f(x)=2x, compute the output of the neuron.

#### OR

- 16. (a) What are ROC space and ROC curve in machine learning? In ROC space, which points correspond to perfect prediction, always positive prediction and always negative prediction? Why?
  - (b) Discuss with a flowchart ,explain how training and testing is performed in backpropagation neural networks? (7)
- 17. (a) Compute the maximum likelihood estimate for the parameter  $\lambda$  in the Poisson (8) distribution whose probability function is  $f(x) = \frac{e^{-\lambda} \lambda^x}{x!}$ 
  - (b) Explain the general MLE method for estimating the parameters of a probability (6) distribution

#### OR

**18.** (a) State the mathematical formulation to express Soft Margin as a constraint optimization problem (8)

- (b) Explain Kernel Trick in the context of support vector machine. List any two kernel function used in SVM.
  (a) Suppose that we have the following data (one variable). Use single linkage Agglomerative clustering to identify the clusters. Data: (2, 5, 9, 15, 16, 18, 25, 33, 33, 45).
  (b) Given two objects represented by the tuples (22, 1, 42, 10) and (20, 0, 36, 8):

  (i) Compute the Euclidean distance between the two objects.
  (ii) Compute the Manhattan distance between the two objects.
  - (iii) Compute the Minkowski distance between the two objects, using p = 3

# OR

- 20. (a) Suppose that we have the following data:
  (2, 0), (1, 2), (2, 2), (3, 2), (2, 3), (3, 3), (2, 4), (3, 4), (4, 4), (3, 5)
  Identify the cluster by applying the k-means algorithm, with k = 2. Try using initial cluster centers as far apart as possible
  - (b) Describe EM algorithm for Gaussian Mixtures

# **Teaching Plan**

No	Contents	
	Module -1 (Overview of machine learning) (8 hours)	
1.1	Introduction to Machine Learning, Machine learning paradigms-supervised, semi-supervised, unsupervised, reinforcement learning.	1 hour
1.2	Supervised learning- Input representation, Hypothesis class, Version space	2 hours
1.3	Vapnik-Chervonenkis (VC) Dimension	2 hours
1.4	Probably Approximately Correct Learning (PAC)	1 hour
1.5	Noise, Learning Multiple classes	1 hour
1.6	Model Selection and Generalization, Overfitting and Underfitting	1 hour

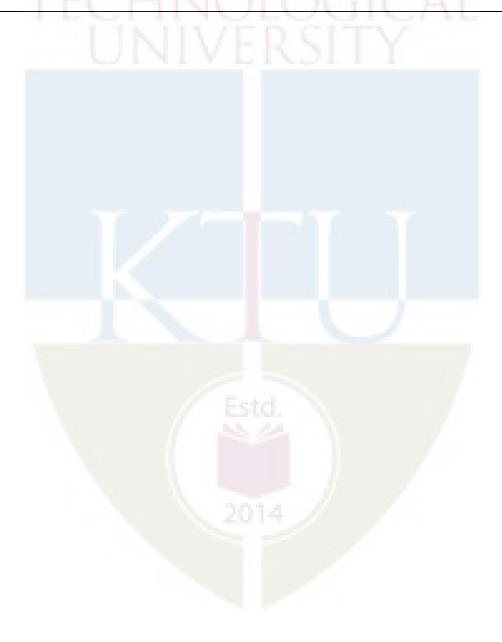
(8)

(6)

	Module-2 (Supervised Learning) (11 hours)	
2.1	Dimensionality reduction – Subset selection, Principal Component Analysis.	2 hours
2.2	Linear regression with one variable (TB 1: Section 2.6)	1 hour
2.3	Multiple variables, Solution using gradient descent algorithm and matrix method (No derivation required) (TB 1: Section 5.8)	2 hours
2.4	Logistic regression	1 hour
2.5	Naive Bayes (TB 2: Section 18.2)	2 hours
2.6	Decision trees (TB 2: Chapter 19)	1 hour
2.7	Decision trees- ID3 algorithm (TB 2: Chapter 19)	1 hour
2.8	Case Study: Develop a classifier for face detection.	1 hour
	Module-3 (Classification Assessment and Neural Networks) (7 hours)	
.1	Performance measures - Precision, Recall, Accuracy, F-Measure, ROC, AUC. (TB 2: Chapter 22.1)	2 hours
3.2	Bootstrapping, Cross validation	1 hour
3.3	Perceptron, Perceptron Learning	1 hour
3.4	Multilayer Feed forward Network, Activation Functions (Sigmoid, ReLU, Tanh)	1 hour
3.5	Back Propagation Algorithm	1 hour
3.6	Illustrative Example for Back Propagation	1 hour
	Module-4 (Parameter estimation & SVM Classifier)) (9 hours)	
4.1	Basics of Parameter estimation	1 hour
4.2	Maximum Likelihood Estimation	1 hour
4.3	Maximum a Posteriori estimation(MAP). Bias-Variance decomposition.	1 hour
4.4	Introduction, Maximum Margin Hyperplane,	1 hour
4.5	Mathematics behind Maximum Margin Classification	1 hour
4.6	Formulation of maximum margin hyperplane and solution	1 hour
4.7	Soft margin SVM, Solution of Soft margin SVM	1 hour
4.8	Non-linear SVM , Kernels for learning non-linear functions, Examples - Linear, RBF, Polynomial, Kernel trick	2 hours
	Module-5 (Unsupervised Learning) (9 hours)	
4.1	Ensemble Methods- Voting, Bagging, Boosting	1 hour

#### B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

4.2	Similarity measures- Minkowski distance measures( Manhattan, Euclidean), Cosine Similarity	1 hour
4.3	K-means clustering (TB 2: Chapter 13)	1 hour
4.4	Clustering - Hierarchical Clustering (TB 2: Chapter 14)	2 hours
4.5	Density based Clustering	2 hours
4.6	Expectation maximization (EM) for soft clustering (TB 2: Chapter 13)	1 hour
4.7	Expectation maximization (EM) for soft clustering (TB 2: Chapter 13)	1 hour
	TECHNOLOGICAL	



AIT 307	AIT 307 INTRODUCTION TO ARTIFICIAL INTELLIGENCE	CATEGORY	L	Т	Р	CREDITS
		PCC	3	1	0	4

# **Preamble:**

The course aims to introduce the fundamental principles of intelligent systems to students. This involves ideas about the characteristics of intelligent systems, knowledge representation schemes, logic and inference mechanisms. The course helps the learner to understand the design of self learning systems along with some of their typical applications in the emerging scenario where the business world is being transformed by the progress made in machine learning.

# **Prerequisite : NIL**

Course Outcomes: After the completion of the course the student will be able to

CO#	СО		
CO1	Explain the fundamental concepts of intelligent systems and their architecture. (Cognitive Knowledge Level: Understanding)		
CO2	Illustrate uninformed and informed search techniques for problem solving in intelligent systems. (Cognitive Knowledge Level: Understanding)		
СО3	Solve Constraint Satisfaction Problems using search techniques. (Cognitive Knowledge Level: Apply )		
CO4	Represent AI domain knowledge using logic systems and use inference techniques for reasoning in intelligent systems. (Cognitive Knowledge Level: Apply)		
CO5	Illustrate different types of learning techniques used in intelligent systems (Cognitive Knowledge Level: Understand)		

	PO 1	PO 2	PO 3	PO 4	PO 5	PO 6	<b>PO</b> 7	PO 8	PO 9	PO1 0	PO1 1	PO1 2
CO 1	$\bigcirc$	Δ	DI	Δ	R I	M	TT	K	ΛT	AN	A	
CO 2		0	-7 -7	î		iñ	10	N		A -	1	Ø
CO 3			0	0	j.	VF	R	ŚŤ	Ťγ	7		٢
CO 4			٢	٢	-			01				
CO 5												

# Mapping of course outcomes with program outcomes

	Abstract POs defined by National Board of Accreditation						
PO #	Broad PO	PO#	Broad PO				
PO1	Engineering Knowledge	PO7	Environment and Sustainability				
PO2	Problem Analysis	PO8	Ethics				
PO3	Design/Development of solutions	PO9	Individual and team work				
PO4	Conduct investigations of complex problems	PO10	Communication				
PO5	Modern tool usage	PO11	Project Management and Finance				
PO6	The Engineer and Society	PO12	Life long learning				

#### **Assessment Pattern**

Bloom's Categor	Contir	nuous Assessment Tests	End Semester Examination Marks (%)
У	Test 1 (%)	Test 2 (%)	
Remember	30	30	30
Understand	60	30	40
Apply	20	40	30
Analyze			

Evaluate		
Create		

#### **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3
	VIII V		CAL-

Continuous Internal Evaluation Pattern:	
Attendance	10 marks
Continuous Assessment Tests(Average of SeriesTests1& 2)	25 marks
Continuous Assessment Assignment	15 marks

### **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed module), having a marks for part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed module), having a marks for part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed module), each with 7 marks. Out of the 7 questions, a student should answer any 5.

## **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 full questions from each module of which student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.

# **Syllabus**

#### Module – 1 (Introduction)

Introduction – What is Artificial Intelligence(AI) ? The Foundations of AI, History of AI, Applications of AI. Intelligent Agents – Agents and Environments, Good behavior: The concept of rationality, Nature of Environments - Specifying the task environment, Properties of task environments. Structure of Agents - Agent programs, Basic kinds of agent programs.

#### Module – 2 (Problem Solving)

Solving Problems by searching-Problem solving Agents, Example problems, Searching for solutions, Uninformed search strategies, Informed search strategies, Heuristic functions.

#### Module - 3 (Search in Complex environments)

Adversarial search - Games, Optimal decisions in games, The Minimax algorithm, Alpha-Beta pruning. Constraint Satisfaction Problems – Defining CSP, Example Problems, Constraint Propagation- inference in CSPs, Backtracking search for CSPs, Structure of CSP problems.

#### Module - 4 (Knowledge Representation and Reasoning)

Logical Agents – Knowledge based agents, Logic, Propositional Logic, Propositional Theorem proving, Agents based on Propositional Logic. First Order Predicate Logic - Syntax and Semantics of First Order Logic, Using First Order Logic, Knowledge representation in First Order Logic. Inference in First Order Logic – Propositional Vs First Order inference, Unification and Lifting, Forward chaining, Backward chaining, Resolution. Classical Planning - Algorithms for planning state space search, Planning Graphs.

#### Module - 5 (Machine Learning)

Learning from Examples – Forms of Learning, Supervised Learning. Learning Decision Trees-The decision tree representation, Inducing decision trees from examples, Choosing attribute tests, Generalization and overfitting. Evaluating and choosing the best hypothesis, Regression and classification with Linear models.

#### **Text Book**

1. Stuart Russell and Peter Norvig. Artificial Intelligence: A Modern Approach, 3rd Edition. Prentice Hall.

#### References

1. Nilsson N.J., Artificial Intelligence - A New Synthesis, Harcourt Asia Pvt. Ltd.

## **Course Level Assessment Questions**

#### **Course Outcome 1 (CO1):**

- 1. Explain about the basic types of agent programs in intelligent systems.
- 2. For the following activities, give a PEAS description of the task environment and characterize it in terms of the task environment properties.
  - a) Playing soccer.
  - b) Bidding on an item at an auction.

### **Course Outcome 2 (CO2):**

- 1. Differentiate between uninformed and informed search strategies in intelligent systems.
- 2. Illustrate the working of Minimax search procedure.

## Course Outcome 3 (CO3):

1. Solve the following crypt arithmetic problem by hand, using the strategy of backtracking with forward checking and the MRV & least-T W Oconstraining-value heuristics. + T W O

#### **Course Outcome 4 (CO4):**

1. Prove, or find a counter example to, the following assertion:

If  $\alpha \models \gamma$  or  $\beta \models \gamma$  (or both) then  $(\alpha \land \beta) \models \gamma$ 

- 2. For each pair of atomic sentences, find the most general unifier if it exists:
- a) P (A, B, B), P (x, y, z).
- b) Q(y, G(A, B)), Q(G(x, x), y).

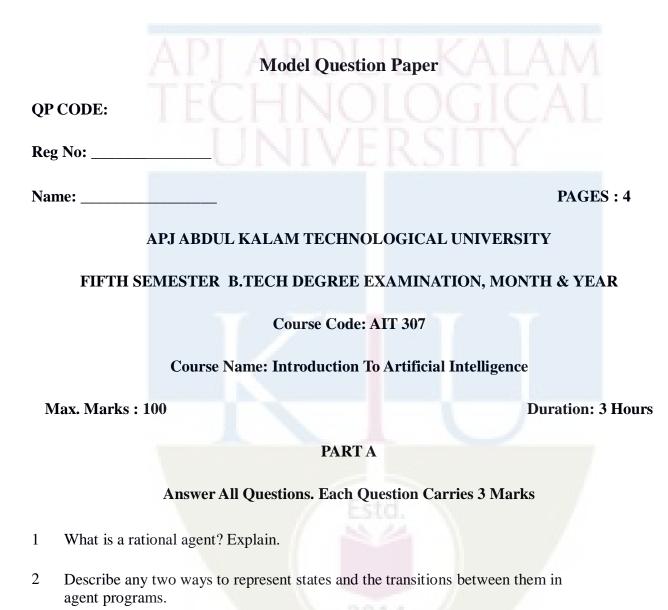
#### **Course Outcome 5 (CO5):**

1. Consider the following data set comprised of three binary input attributes (A1, A2, and A3) and one binary output.

Example	$A_1$	$A_2$	$A_3$	Output $y$
<b>X</b> 1	1	0	0	0
$\mathbf{x}_2$	1	0	1	0
$\mathbf{x}_3$	0	1	0	0
$\mathbf{x}_4$	1	1	1	1
$\mathbf{x}_5$	1	1	0	1

Use the DECISION-TREE-LEARNING algorithm to learn a decision tree for these data. Show the computations made to determine the attribute to split at each node.

2. What is multivariate linear regression? Explain.



- 2014
- 3 Differentiate between informed search and uninformed search.
- 4 Define heuristic function? Give two examples.
- 5 What are the components of a Constraint Satisfaction Problem? Illustrate with an example.
- 6 Formulate the following problem as a CSP. Class scheduling: There is a fixed number of professors and classrooms, a list of classes to be offered, and a list of possible time slots for classes. Each professor has a set of classes that he or she can teach.

(10x3=30)

- 7 What is a knowledge based agent? How does it work?
- 8. Represent the following assertion in propositional logic:"A person who is radical (R) is electable (E) if he/she is conservative (C), but otherwise is not electable."
- 9 Describe the various forms of learning?
- 10 State and explain Ockham's razor principle?

#### Part B

## (Answer any one question from each module. Each question carries 14 Marks)

- 11. (a) Explain the structure Goal-based agents and Utility-based agents with the help of diagrams. (8)
  - (b) For the following activities, give a PEAS description of the task environment (6) and characterize it in terms of the task environment properties.
    - a) Playing soccer
    - b) Bidding on an item at an auction.

#### OR

12. (a) Explain the structure of Simple reflex agents and Model-based reflex agents with the help of diagrams.
(b) Discuss about any five applications of AI.
(c) 13. (a) Explain Best First Search algorithm. How does it implement heuristic search?
(b) Describe any four uninformed search strategies.
(c) (b) Describe any four uninformed search strategies.

#### OR

- 14. (a) Write and explain A\* search algorithm. (6)
  - (b) Explain the components of a well defined AI problem? Write the standard (8) formulation of 8-puzzle problem.

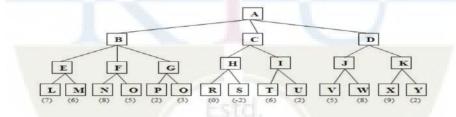
15. (a) (a) Solve the following crypt arithmetic problem by hand, using the strategy (8) of backtracking with forward checking and the MRV and least-constraining-value heuristics.



(b) What is local consistency in CSP constraint propagation? Explain different types of local consistencies.

#### OR

- 16. (a) Illustrate the use of alpha-beta pruning in games.
  - (b) Consider the following game tree in which static evaluation score are all from (8) the players point of view: static evaluation score range is (+10 to -10)



Suppose the first player is the maximizing player. What move should be chosen? Justify your answer.

17. (a) Convert the following sentences into first order logic:

Everyone who loves all animals is loved by someone.

Anyone who kills an animal is loved by no one.

Jack loves all animals.

Either Jack or Curiosity killed the cat, who is named Tuna.

Did Curiosity kill the cat?

(b) Give a resolution proof to answer the question "Did Curiosity kill the cat?" (8)

(6)

(6)

(6)

## OR

18.	(a)	Draw a plevel S2	planning graph for the "have cake and eat cake too" problem up to	(6)
	(b)		n pair of atomic sentences, give the most general unifier if it exists: Father (y), y), Older (Father (x), John).	(8)
19.	(a)	How is	best hypothesis selected from alternatives?	(8)
	(b)	Explain	Univariate Linear Regression.	(6)
			OR	

20. (a) Consider the following data set comprised of two binary input attributes (A1 (8) and A2) and one binary output.

Example	Aı	<b>A</b> <sub>2</sub>	Output y
<b>X</b> 1	1	1	1
X2	1	1	1
X3	1	0	0
X4	0	0	1
Xs	0	1	0
X <sub>6</sub>	0	1	0

Use the DECISION-TREE-LEARNING algorithm to learn a decision tree for these data. Show the computations made to determine the attribute to split at each node.

(b) Explain Linear classification with logistic regression

(6)

# **TEACHING PLAN**

No	Contents	No of Lecture Hrs (44)
	Module – 1 (Introduction) (9 hrs)	
1.1	Introduction, What is Artificial Intelligence(AI)?	1
1.2	The foundations of AI, The history of AI	1
1.3	Applications of AI	1
1.4	Intelligent Agents – Agents and Environments	1
1.5	Good behavior: The concept of rationality	1
1.6	The nature of Environments Specifying the task environment	1
1.7	Properties of task environments	1
1.8	The structure of Agents - Agent programs	1
1.9	Basic kinds of agent programs	1
	Module - 2 (Problem Solving by searching) (7 hrs)	
2.1	Solving Problems by searching-Problem solving Agents	1
2.2	Illustration of the problem solving process by agents	1
2.3	Searching for solutions	1
2.4	Uninformed search strategies:BFS, Uniform-cost search, DFS, Depth-	1
	limited search, Iterative deepening depth-first search	
2.5	Informed search strategies: Best First search	1
2.6	Informed search strategies: A* Search	1
2.7	Heuristic functions	1
	Module - 3 (Problem Solving in complex environments ) (8 hrs)	
3.1	Adversarial search - Games	1
3.2	Optimal decisions in games, The Minimax algorithm	1
3.3	Alpha-Beta pruning	1
3.4	Constraint Satisfaction Problems – Defining CSP	1
3.5	Example Problem formulations	1
3.6	Constraint Propagation- inference in CSPs	1
3.7	Backtracking search for CSPs	1
3.8	The structure of problems	1

	Module - 4 (Knowledge Representation and Reasoning) (12 hrs)	
4.1	Logical Agents – Knowledge based agents and logic	1
4.2	Propositional Logic	1
4.3	Propositional Theorem proving	1
4.4	Agents based on Propositional Logic	1

	First Order Predicate Logic – Syntax and Semantics of First Order	1
	Logic	
4.6	Using First Order Logic, Knowledge representation in First Order	1
	Logic	
4.7	Inference in First Order Logic – Propositional Vs First Order inference,	1
	Unification and Lifting	A.
4.8	Forward chaining, Backward chaining	1
4.9	Resolution	1
4.10	Classical Planning	1
4.11	Algorithms for planning state space search	1
4.12	Planning Graphs	1

5.1	Learning from Examples – Forms of Learning	1
5.2	Supervised Learning	1
5.3	Learning Decision Trees- The decision tree representation	1
5.4	Inducing decision trees from examples	1
5.5	Choosing attribute tests	1
5.6	Generalization and overfitting	1
5.7	Evaluating and choosing the best hypothesis	1
5.8	Regression and classification with Linear models.	1



CST	MANAGEMENT OF	Category	L	Т	Р	Credit	Year of Introduction
309	SOFTWARE SYSTEMS	PCC	3	0	0	3	2019

**Preamble**: This course provides fundamental knowledge in the Software Development Process. It covers Software Development, Quality Assurance, Project Management concepts and technology trends. This course enables the learners to apply state of the art industry practices in Software development.

Prerequisite: Basic understanding of Object Oriented Design and Development.

**Course Outcomes**: After the completion of the course the student will be able to

CO1	Demonstrate Traditional and Agile Software Development approaches (Cognitive
COI	Knowledge Level: Apply)
CO2	Prepare Software Requirement Specification and Software Design for a given
002	problem. (Cognitive Knowledge Level: Apply)
	Justify the significance of design patterns and licensing terms in software
CO3	development, prepare testing, maintenance and DevOps strategies for a project.
	(Cognitive Knowledge Level: Apply)
	Make use of software project management concepts while planning, estimation,
CO4	scheduling, tracking and change management of a project, with a traditional/agile
	framework. (Cognitive Knowledge Level: Apply)
CO5	Utilize SQA practices, Process Improvement techniques and Technology
COS	advancements in cloud based software models and containers & microservices.
	(Cognitive Knowledge Level: Apply)

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	9	9	9	D	9		Á	LA	M		9
CO2		9		9	Ń	9	Q		Ģ	9	9	9
CO3			9	0	Y	tł	0		Y	9	9	9
CO4	9	0	9			9					9	9
CO5						9						9

# Mapping of course outcomes with program outcomes

	Abstract POs defined by National Board of Accreditation							
PO#	Broad PO	PO#	Broad PO					
PO1	Engineering Knowledge	P07	Environment and Sustainability					
PO2	Problem Analysis	PO8	Ethics					
PO3	Design/Development of solutions	PO9	Individual and team work					
PO4	Conduct investigations of complex problems	PO10	Communication					
PO5	Modern tool usage	PO11	Project Management and Finance					
PO6	The Engineer and Society	PO12	Lifelong learning					

## **Assessment Pattern**

Plaam's Catagowy	Continuous Asses	End Semester		
Bloom's Category	Test1 (Percentage)	Test2 (Percentage)	Examination Marks	
Remember	30	30	30	
Understand	40	40	50	
Apply	30	30	20	
Analyse	UTATAT	NOLL I		
Evaluate				
Create				

## **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3 hours

## **Continuous Internal Evaluation Pattern:**

Attendance: 10 marksContinuous Assessment Tests: 25 marks

Continuous Assessment Assignment : **15 marks** (Each student shall identify a software development problem and prepare Requirements Specification, Design Document, Project Plan and Test case documents for the identified problem as the assignment.)

## **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks.

First Internal Examination shall be preferably conducted after completing the first half of the syllabus and the Second Internal Examination shall be preferably conducted after completing the remaining part of the syllabus.

There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly covered module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly covered module), each with 7 marks. Out of the 7 questions in Part B, a student should answer any 5.

#### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have a maximum of 2 subdivisions and carries 14 marks.

#### **Syllabus**

#### Module 1 : Introduction to Software Engineering (7 hours)

Introduction to Software Engineering - Professional software development, Software engineering ethics. Software process models - The waterfall model, Incremental development. Process activities - Software specification, Software design and implementation, Software validation, Software evolution. Coping with change - Prototyping, Incremental delivery, Boehm's Spiral Model. Agile software development - Agile methods, agile manifesto - values and principles. Agile development techniques, Agile Project Management. Case studies: An insulin pump control system. Mentcare - a patient information system for mental health care.

## Module 2 : Requirement Analysis and Design (8 hours)

Functional and non-functional requirements, Requirements engineering processes. Requirements elicitation, Requirements validation, Requirements change, Traceability Matrix. Developing use cases, Software Requirements Specification Template, Personas, Scenarios, User stories, Feature identification. Design concepts - Design within the context of software engineering, Design Process, Design concepts, Design Model. Architectural Design - Software Architecture, Architectural Styles, Architectural considerations, Architectural Design Component level design - What is a component?, Designing Class-Based Components, Conducting Component level design, Component level design for web-apps. Template of a Design Document as per "IEEE Std 1016-2009 IEEE Standard for Information Technology Systems Design Software Design Descriptions". Case study: The Ariane 5 launcher failure.

## Module 3 : Implementation and Testing (9 hours)

Object-oriented design using the UML, Design patterns, Implementation issues, Open-source development - Open-source licensing - GPL, LGPL, BSD. Review Techniques - Cost impact of Software Defects, Code review and statistical analysis. Informal Review, Formal Technical Reviews, Post-mortem evaluations. Software testing strategies - Unit Testing, Integration Testing, Validation testing, System testing, Debugging, White box testing, Path testing, Control Structure testing, Black box testing, Testing Documentation and Help facilities. Test automation, Test-driven development, Security testing. Overview of DevOps and Code Management - Code management, DevOps automation, Continuous Integration, Delivery, and Deployment (CI/CD/CD). Software Evolution - Evolution processes, Software maintenance.

## Module 4 : Software Project Management (6 hours)

Software Project Management - Risk management, Managing people, Teamwork. Project Planning, Software pricing, Plan-driven development, Project scheduling, Agile planning. Estimation techniques, COCOMO cost modeling. Configuration management, Version management, System building, Change management, Release management, Agile software management - SCRUM framework. Kanban methodology and lean approaches.

## Module 5 : Software Quality, Process Improvement and Technology trends (6 hours)

Software Quality, Software Quality Dilemma, Achieving Software Quality Elements of Software Quality Assurance, SQA Tasks, Software measurement and metrics. Software Process Improvement(SPI), SPI Process CMMI process improvement framework, ISO 9001:2000 for Software. Cloud-based Software - Virtualisation and containers, Everything as a service(IaaS, PaaS), Software as a service. Microservices Architecture - Microservices, Microservices architecture, Microservice deployment.

## **Text Books**

- 1. Book 1 Ian Sommerville, Software Engineering, Pearson Education, Tenth edition, 2015.
- 2. Book 2 Roger S. Pressman, Software Engineering : A practitioner's approach, McGraw Hill publication, Eighth edition, 2014
- 3. Book 3 Ian Sommerville, Engineering Software Products: An Introduction to Modern Software Engineering, Pearson Education, First Edition, 2020.

#### References

- 1. IEEE Std 830-1998 IEEE Recommended Practice for Software Requirements SpeciPcations
- 2. IEEE Std 1016-2009 IEEE Standard for Information Technology—Systems Design— Software Design Descriptions

- 3. David J. Anderson, Kanban, Blue Hole Press 2010
- 4. David J. Anderson, Agile Management for Software Engineering, Pearson, 2003
- 5. Walker Royce, Software Project Management : A unified framework, Pearson Education, 1998
- 6. Steve. Denning, The age of agile, how smart companies are transforming the way work gets done. New York, Amacom, 2018.
- 7. Satya Nadella, Hit Refresh: The Quest to Rediscover Microsoft's Soul and Imagine a Better Future for Everyone, Harper Business, 2017
- 8. Henrico Dolfing, Project Failure Case Studies: Lessons learned from other people's mistakes, Kindle edition
- 9. Mary Poppendieck, Implementing Lean Software Development: From Concept to Cash, Addison-Wesley Signature Series, 2006
- 10. StarUML documentation https://docs.staruml.io/
- 11. OpenProject documentation https://docs.openproject.org/
- 12. BugZilla documentation https://www.bugzilla.org/docs/
- 13. GitHub documentation https://guides.github.com/
- 14. Jira documentation https://www.atlassian.com/software/jira

## **Course Level Assessment Questions**

## **Course Outcome 1 (CO1):**

- 1. What are the advantages of an incremental development model over a waterfall model?
- 2. Illustrate how the process differs in agile software development and traditional software development with a socially relevant case study. (Assignment question)

## Course Outcome 2 (CO2):

- 1. How to prepare a software requirement specification?
- 2. Differentiate between Architectural design and Component level design.
- 3. How does agile approaches help software developers to capture and define the user requirements effectively?
- 4. What is the relevance of the SRS specification in software development?
- 5. Prepare a use case diagram for a library management system.

# Course Outcome 3 (CO3):

- 1. Differentiate between the different types of software testing strategies.
- 2. Justify the need for DevOps practices?
- 3. How do design patterns help software architects communicate the design of a complex system effectively?

4. What are the proactive approaches one can take to optimise efforts in the testing phase?

## **Course Outcome 4 (CO4):**

- 1. Illustrate the activities involved in software project management for a socially relevant problem?
- 2. How do SCRUM, Kanban and Lean methodologies help software project management?
- 3. Is rolling level planning in software project management beneficial? Justify your answer.
- 4. How would you assess the risks in your software development project? Explain how you can manage identified risks?

## **Course Outcome 5 (CO5):**

- 1. Justify the importance of Software Process improvement?
- 2. Explain the benefits of cloud based software development, containers and microservices.
- 3. Give the role of retrospectives in improving the software development process.
- 4. Illustrate the use of project history data as a prediction tool to plan future socially relevant projects.



## **Model Question Paper**

QP CODE:	
Reg No:	
Name :	PAGES : 3
APJ ABDUL KALAM TECHNOLO	OGICAL UNIVERSITY
FIFTH SEMESTER B.TECH DEGREE EXA	AMINATION, MONTH & YEAR
Course Code: CST	309
Course Name: Management of	f Software Systems

#### **Duration: 3 Hrs**

#### Max. Marks :100

## PART A

## Answer all Questions. Each question carries 3 marks

- 1. Why professional software that is developed for a customer is not simply theprograms that have been developed and delivered.
- 2. Incremental software development could be very effectively used for customers who do not have a clear idea about the systems needed for their operations. Justify.
- 3. Identify any four types of requirements that may be defined for a software system
- 4. Describe software architecture
- Differentiate between GPL and LGPL? 5.
- Compare white box testing and black box testing. 6.
- 7. Specify the importance of risk management in software project management?
- 8. Describe COCOMO cost estimation model.
- 9. Discuss the software quality dilemma
- 10. List the levels of the CMMI model?

# Part B

## (Answer any one question from each module. Each question carries 14 Marks)

11. (a) Compare waterfall model and spiral model

(10x3=30))

## B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

	(b)	Explain Agile ceremonies and Agile manifesto	(6)
12.	(a)	Illustrate software process activities with an example.	(8)
	(b)	Explain Agile Development techniques and Agile Project Management	(6)
13.	(a)	What are functional and nonfunctional requirements? Imagine that you are developing a library management software for your college, list eight functional requirements and four nonfunctional requirements.	(10)
	(b)	List the components of a software requirement specification?	(4)
		OR	
14.	(a)	Explain Personas, Scenarios, User stories and Feature identification?	(8)
	(b)	Compare Software Architecture design and Component level design	(6)
15.	(a)	Explain software testing strategies.	(8)
	(b)	Describe the formal and informal review techniques.	(6)
		OR	
16.	(a)	Explain Continuous Integration, Delivery, and Deployment CI/CD/CD)	
	<b>4</b> \		(8)
	(b)	Explain test driven development	(6)
17.	(a)	What is a critical path and demonstrate its significance in a project schedule with the help of a sample project schedule.	(8)
	(b)	Explain plan driven development and project scheduling.	(6)
		OR	
18.	(a)	Explain elements of Software Quality Assurance and SQA Tasks.	(6)
	(b)	What is algorithmic cost modeling? What problems does it suffer from when	(8)

compared with other approaches to cost estimation?

19. (a)	Explain elements of Software Quality Assurance and SQA Tasks.	(8)
(b)	Illustrate SPI process with an example.	(6)
	ALLADUCOR ALAM	
20. (a)	Compare CMMI and ISO 9001:2000.	(8)
(b)	How can Software projects benefit from Container deployment and Micro service deployment?	(6)

# **Teaching Plan**

No	Contents	No of Lecture Hrs
	Module 1 : Introduction to Software Engineering (7 hours)	
1.1	Introduction to Software Engineering. [Book 1, Chapter 1]	1 hour
1.2	Software process models [Book 1 - Chapter 2]	1 hour
1.3	Process activities [Book 1 - Chapter 2]	1 hour
1.4	Coping with change [Book 1 - Chapter 2, Book 2 - Chapter 4]	1 hour
1.5	Case studies : An insulin pump control system. Mentcare - a patient information system for mental health care. [Book 1 - Chapter 1]	1 hour
1.6	Agile software development [Book 1 - Chapter 3]	1 hour
1.7	Agile development techniques, Agile Project Management.[Book 1 - Chapter 3]	1 hour
	Module 2 : Requirement Analysis and Design (8 hours)	
2.1	Functional and non-functional requirements, Requirements engineering processes [Book 1 - Chapter 4]	1 hour
2.2	Requirements elicitation, Requirements validation, Requirements change, Traceability Matrix [Book 1 - Chapter 4]	1 hour
2.3	Developing use cases, Software Requirements Specification Template [Book 2 - Chapter 8]	1 hour

2.4	Personas, Scenarios, User stories, Feature identification [Book 3 - Chapter 3]	1 hour
2.5	Design concepts [Book 2 - Chapter 12]	1 hour
2.6	Architectural Design [Book 2 - Chapter 13]	1 hour
2.7	Component level design [Book 2 - Chapter 14]	1 hour
2.8	Design Document Template. Case study: The Ariane 5 launcher failure. [Ref	1 hour
	- 2, Book 2 - Chapter 16]	
	Module 3 : Implementation and Testing (9 hours)	
3.1	Object-oriented design using the UML, Design patterns [Book 1 - Chapter 7]	1 hour
3.2	Implementation issues, Open-source development - Open-source licensing -GPL, LGPL, BSD [Book 1 - Chapter 7]	1 hour
3.3	Review Techniques - Cost impact of Software Defects, Code review andstatistical analysis. [Book 2 - Chapter 20]	1 hour
34	Informal Review, Formal Technical Reviews, Post-mortem evaluations.[Book 2 - Chapter 20]	1 hour
3.5	Software testing strategies - Unit Testing, Integration Testing, Validation testing, System testing and Debugging (basic concepts only). [Book 2 - Chapter 22]	1 hour
3.6	White box testing, Path testing, Control Structure testing, Black box testing. Test documentation [Book 2 - Chapter 23]	1 hour
3.7	Test automation, Test-driven development, Security testing. [Book 3 -Chapter 9]	1 hour
3.8	DevOps and Code Management - Code management, DevOps automation,CI/CD/CD. [Book 3 - Chapter 10]	1 hour
3.9	Software Evolution - Evolution processes, Software maintenance. [Book 1 -Chapter 9]	1 hour
	Module 4 : Software Project Management (6 hours)	
4.1	Software Project Management - Risk management, Managing people, Teamwork [Book 1 - Chapter 22]	1 hour
4.2	Project Planning - Software pricing, Plan-driven development, Project scheduling, Agile planning [Book 1 - Chapter 23]	1 hour
4.3	Estimation techniques [Book 1 - Chapter 23]	1 hour
4.4	Configuration management [Book 1 - Chapter 25]	1 hour

## B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

4.5	Agile software management - SCRUM framework [Book 2 - Chapter 5]	1 hour
4.6	Kanban methodology and lean approaches.[Ref 9 - Chapter 2]	1 hour
M	odule 5 : Software Quality, Process Improvement and Technology trends ( hours)	6
5.1	Software Quality, Software Quality Dilemma, Achieving Software Quality. [Book 2 - Chapter 19]	1 hour
5.2	Elements of Software Quality Assurance, SQA Tasks, Software measurement and metrics. [Book 3 - Chapter 21]	1 hour
5.3	Software Process Improvement (SPI), SPI Process [Book 2 - Chapter 37]	1 hour
5.4	CMMI process improvement framework, ISO 9001:2000 for Software. [Book 2 - Chapter 37]	1 hour
5.5	Cloud-based Software - Virtualisation and containers, IaaS, PaaS, SaaS.[Book 3 - Chapter 5]	1 hour
5.6	Microservices Architecture - Microservices, Microservices architecture, Microservice deployment [Book 3 - Chapter 6]	1 hour



AML 331	PYTHON AND MACHINE LEARNING	CATEGORY	L	Т	Р	Credit	Year of Introduction
	LAB	LAB	0	0	3	2	2022

## **Preamble**:

This course enables the learners to get hands-on experience in most popular supervised learning algorithms (such as linear regression, logistic regression, decision trees, Bayesian learning and Naive Bayes algorithm) and unsupervised learning algorithms (such as basic clustering algorithms). This helps the learners to understand the process of knowledge inference from raw data through dataset preprocessing and analysis.

**Prerequisite**: Fundamentals of Programming, Python programming fundamentals, Machine learning.

Course Outcomes: After the completion of the course the student will be able to

CO#	Course Outcomes
CO1	Develop applications in Python programming. (Cognitive Knowledge Level: Apply)
CO2	Implement machine learning algorithms using packages and libraries in Python for various applications.(Cognitive Knowledge Level: Apply)
CO3	Implement python programs for supervised learning methods through Neural network, Regression and classification.(Cognitive Knowledge Level: Apply)
CO4	Implement clustering algorithms.(Cognitive Knowledge Level: Apply)
CO5	Apply dimensionality reduction as a dataset preprocessing step. (Cognitive Knowledge Level: Apply)

## Mapping of course outcomes with program outcomes

	PO1	PO2	PO3	PO4	PO5	PO6	<b>PO7</b>	PO8	PO9	PO10	PO11	PO12
CO1	$\oslash$	$\oslash$	$\oslash$	$\oslash$				$\oslash$				$\oslash$
CO2	$\oslash$	$\oslash$	$\oslash$	$\oslash$	$\oslash$			$\oslash$				$\oslash$
CO3	$\oslash$	$\oslash$	$\oslash$	$\oslash$	$\oslash$	$\bigcirc$		$\oslash$				$\oslash$
CO4	$\oslash$	$\oslash$	$\oslash$	$\oslash$	$\oslash$	$\oslash$		$\oslash$				$\oslash$

#### B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

CO5												$\bigotimes$
-----	--	--	--	--	--	--	--	--	--	--	--	--------------

Abstr	act POs defined by National Board of Accredi	tation	CALAM			
PO#	Broad PO	PO#	Broad PO			
PO1	Engineering Knowledge	PO7	Environment and Sustainability			
PO2	Problem Analysis	PO8	Ethics			
PO3	Design/Development of solutions	PO9	Individual and teamwork			
PO4	Conduct investigations of complex problems	PO10	Communication			
PO5	Modern tool usage	PO11	Project Management and Finance			
PO6	The Engineer and Society	PO12	Lifelong learning			

# Assessment Pattern

Bloom's Category	Continuous Assessment Test (Internal Exam) Marks in percentage	End Examination Marks in perce	Semester
Remember	20	20	-
Understand	20	20	)
Apply	60	60	)
Analyze	Estd.		
Evaluate		1	
Create			

## Mark Distribution

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	75	75	3 hours

## **Continuous Internal Evaluation Pattern:**

Attendance	15 marks
Continuous Evaluation in Lab	30 marks
Continuous Assessment Test	15 marks
Viva voce	15 marks

## **Internal Examination Pattern:**

The Internal examination shall be conducted for 100 marks, which will be converted to out of 15 while calculating Internal Evaluation marks. The marks will be distributed as, Algorithm - 30 marks, Program - 20 marks, Output - 20 marks and Viva - 30 marks.

## **End Semester Examination Pattern:**

The End Semester Examination shall be conducted for 100 marks, which will be converted to out of 75 while calculating External Evaluation marks. The marks will be distributed as, Algorithm - 30 marks, Program - 20 marks, Output - 20 marks and Viva- 30 marks.

## **Operating System to use in lab : Ubuntu , Windows**

#### Fair Lab Record:

All the students attending the Python and machine learning lab should have a Fair Record. Every experiment conducted in the lab should be noted in the fair record. For every experiment, in the fair record, the right-hand page should contain experiment heading, experiment number, date of experiment, aim of the experiment, procedure/algorithm followed, other such details of the experiment and final result. The left-hand page should contain a print out of the respective code with sample input and corresponding output obtained. All the experiments noted in the fair record should be verified by the faculty regularly. The fair record, properly certified by the faculty, should be produced during the time of End Semester Examination for the verification by the examiners.

# **Syllabus**

## \*Mandatory

1. Introduction to Python Programming.

- 2. Familiarization of basic Python Libraries such as Sklearn, Numpy, Pandas and Matplotlib.\*
- 3. Write a Python program to find the union and intersection of two lists.
- 4. Design a Python program to count the occurrences of each word in a given sentence.
- 5. Write a Python program to multiply two matrices.\*
- 6. Write a Python program to find the most frequent words in a text file.\*
- 7. Implement and demonstrate Single, Multi variable and Polynomial Regression for a given set of training data stored in a .CSV file and evaluate the accuracy.\*
- 8. Implement a Python program to perform logistic regression on a dataset.
- 9. Write a Python program to implement Naive Bayes classifier and calculate the accuracy, precision, and recall for your data set.\*
- 10. Write a Python program to demonstrate the working of the decision tree based ID3 algorithm. Use an appropriate data set for building the decision tree and apply this knowledge to classify a new sample.\*
- 11. Assuming a set of data that need to be classified, use a Support Vector Machine classifier to perform this task and evaluate the accuracy.\*
- 12. Implement K-Nearest Neighbor algorithm to classify any dataset.
- 13. Implement K-Means Clustering using any given dataset.\*
- 14. Build an Artificial Neural Network using Backpropagation algorithm and test the same with appropriate dataset.\*
- 15. Implement dimensionality reduction using PCA.

(Use socially relevant dataset as far as possible)

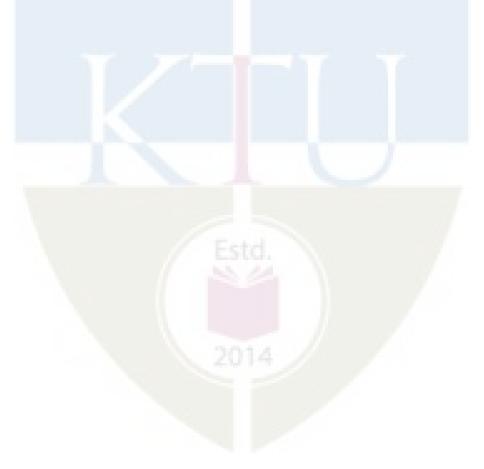
# **Python and Machine Learning Lab-Practice Questions**

- 1. Review of Python programming: Programs using matplotlib / plotly / bokeh / seaborn.
- 2. Write a program to find words which are greater than a given length k.
- 3. Write a program to construct a Bayesian network considering medical data. Use this model to demonstrate the diagnosis of heart patients using the standard Heart Disease Dataset.
- 4. Implementation of Random Forest Classification in Python.
- 5. Write a program to implement k-Nearest Neighbor algorithm to classify the iris data set. Print both correct and wrong predictions.
- 6. Program to implement text classification using Support vector machine.
- 7. The probability that it is Friday and that a student is absent is 3%. Since there are 5 school days in a week, the probability that it is Friday is 20%. What is the probability that a student is absent given that today is Friday? Apply Bayes' rule in python to get the result. (Ans: 15%)

- 8. Implement Naïve Bayes theorem to classify the English text.
- 9. Program to implement Mean-Shift algorithm in python.
- 10. Implement Agglomerative Hierarchical Clustering.
- 11. Apply K-Means clustering to evaluate Student's performance. The results expected show the profile of a student with criteria for excellent performance, standard performance, and underperformance.

## **Reference Books:**

- 1. Aurelien Geron, "Hands-On Machine Learning with Scikit-Learn and TensorFlow", O'Relily.
- 2. Christopher Bishop, "Pattern Recognition and Machine Learning", Springer, 2007.
- 3. Ian Good fellow and Yoshua Bengio and Aaron Courville, "Deep Learning", MIT Press, 2016.



AIL 333	AI ALGORITHMS LAB	CATEGORY	L	Т	Р	Credit	Year of Introduction
	ADI /	LAB	0	0	3	2	2022
	AL L	TDLAC	1		17	<u></u>	VIA1

## **Preamble**:

This laboratory course enables the students to get the fundamental concepts in the area of Artificial Intelligence. This course covers the AI based Algorithms, logical reasoning agents and implementation of these reasoning systems using either backward or forward inference mechanisms. This course helps the learners to apply AI techniques to solve real world problems.

**Prerequisite**: A sound knowledge of the basics of programming, Discrete Mathematics.

**Course Outcomes**: After the completion of the course, the student will be able to:

CO#	Course Outcomes
CO1	State the basics of learning problems with hypothesis and version spaces (Cognitive Knowledge Level: Understand).
CO2	Demonstrate real-world problems as state space problems, optimization problems or constraint satisfaction problems. (Cognitive Knowledge Level: Apply)
CO3	Simulate given problem scenario and analyze its performance. (Cognitive Knowledge Level: Apply)
CO4	Develop programming solutions for given problem scenario. (Cognitive Knowledge Level: Apply)
CO5	Design and develop an expert system by using appropriate tools and techniques. (Cognitive Knowledge Level: Apply)

	PO1	PO2	PO3	PO4	PO5	PO6	<b>PO7</b>	PO8	PO9	<b>PO10</b>	PO11	PO12
CO1	$\oslash$	$\bigcirc$	$\bigcirc$	P.P.	17	2	1	0	5	ΠÂI		$\bigcirc$
CO2	$\oslash$	$\bigcirc$	$\bigcirc$	0	N	DI.	0	0	6	AL		$\oslash$
CO3	$\oslash$	$\oslash$	$\oslash$	$\bigcirc$	0	E	tS	$\oslash$	Y			$\oslash$
CO4	0	$\oslash$	$\bigcirc$	$\oslash$	$\bigcirc$	$\bigcirc$		$\oslash$				$\oslash$
CO5	Ø	$\oslash$	$\oslash$	$\oslash$	$\oslash$			$\oslash$				

## Mapping of course outcomes with program outcomes

Abstra	n		
PO#	Broad PO	PO#	Broad PO
PO1	Engineering Knowledge	<b>PO7</b>	Environment and Sustainability
PO2	Problem Analysis	<b>PO8</b>	Ethics
PO3	Design/Development of solutions	PO9	Individual and teamwork
PO4	Conduct investigations of complex problems	PO10	Communication
PO5	Modern tool usage	PO11	Project Management and Finance
PO6	The Engineer and Society	PO12	Lifelong learning

# **Assessment Pattern**

Bloom's Category	Continuous Assessment Test (Internal Exam) Marks in percentage	End Semester Examination Marks in percentage
Remember	20	20
Understand	20	20
Apply	60	60
Analyze		
Evaluate		

#### B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

Create	

#### **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	75	75	- 3 hours

#### **Continuous Internal Evaluation Pattern:**

Attendance	: 15 marks
Continuous Evaluation in Lab	: 30 marks
Continuous Assessment Test	: 15 marks
Viva voce	: 15 marks

## **Internal Examination Pattern:**

The internal examination shall be conducted for 100 marks, which will be converted to out of 15, while calculating internal evaluation marks. The marks will be distributed as, Algorithm - 30 marks, Program - 20 marks, Output - 20 marks and Viva - 30 marks.

#### **End Semester Examination Pattern:**

The end semester examination will be conducted for a total of 75 marks and shall be distributed as, Algorithm - 30 marks, Program - 20 marks, Output - 20 marks and Viva- 30 marks.

Operating System to Use in Lab	: Linux/Windows
Programming Language to Use in Lab	: C++/Java/Python/Prolog

#### Fair Lab Record:

All the students attending the Artificial Intelligence Algorithms laboratory should have a fair record. Every experiment conducted in the lab should be noted in the fair record. For every experiment, in the fair record, the right-hand page should contain experiment heading, experiment number, date of experiment, aim of the experiment, procedure/algorithm followed, other such details of the experiment and final result. The left-hand page should contain a print out of the respective code with sample input and corresponding output obtained. All the experiments noted in the fair record should be verified by the faculty

regularly. The fair record, properly certified by the faculty, should be produced during the time of end semester examination for the verification by the examiners.

# Syllabus

## \*Mandatory

- 1. Installation and working on various AI tools viz. Python, R, GATE, NLTK, MATLAB etc.\*
- 2. Implement basic search strategies for selected AI applications\*.
- 3. Implement state space search algorithms\*
- 4. Implement informed search algorithms\*
- 5. Implement backtracking algorithms for CSP\*
- 6. Implement local search algorithms for CSP\*
- 7. Implement propositional logic inferences for AI tasks\*
- 8. Implementation of Knowledge representation schemes\*
- 9. Implement travelling salesman problem\*
- 10. Implementation of Game playing (adversarial search)
- 11. Mini Project that implement a real world application using AI techniques (Group project with a maximum of four students)

## **PRACTICE QUESTIONS**

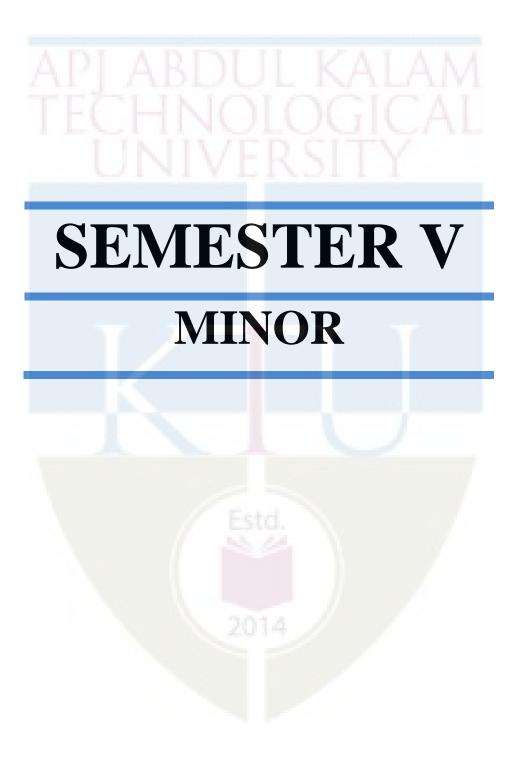
- 1. Implementation of Depth-First Search (DFS).
- 2. Write a program to implement water jug problem.
- 3. Implement variants of hill-climbing and genetic algorithms.
- 4. Implement tic tac toe game for 0 and X.
- 5. Develop a program to construct a pruned game tree using Alpha-Beta pruning. Take the sequence, [5, 3, 2, 4, 1, 3, 6, 2, 8, 7, 5, 1, 3, 4] of MINIMAX values for the nodes at the cutoff depth of 4 plies. Assume that branching factor is 2, MIN makes the first move, and nodes are generated from right to left.
- 6. Write a program to implement production system.

- 7. Write a program to implement heuristic search procedure.
- 8. Write a program to implement Expert system.
- 9. Write a program to implement search problem of 3 x 3 puzzles.

#### **References:**

- 1. Dan W. Patterson, "Introduction to AI and ES", Pearson Education, 2007
- 2. Kevin Night, Elaine Rich, and Nair B., "Artificial Intelligence", McGraw Hill, 2008
- 3. Patrick H. Winston, "Artificial Intelligence", Third edition, Pearson Edition, 2006
- 4. Deepak Khemani, "Artificial Intelligence", Tata McGraw Hill Education, 2013 (http://nptel.ac.in/)
- 5. Artificial Intelligence by Example: Develop machine intelligence from scratch using real artificial intelligence use cases -by Dennis Rothman, 2018
- 6. Padhy, N.P. 2009. Artificial Intelligence and Intelligent Systems, Oxford University Press
- 7. Brachman, R. and Levesque, H. 2004. Knowledge Representation and Reasoning, Morgan Kaufmann.





CS7 381	CONCEPTS IN SOFTWARE ENGINEERING			Т	Р	Credit	Year of Introduction
	ENGINEERING	VAC	3	1	0	4 A 1 /	2019
	ALLAL	1.			P		VIVI.

**Preamble**: This course provides fundamental knowledge in the Software Development Process. It covers Software Development, Quality Assurance and Project Management concepts. This course enables the learners to apply state of the art industry practices in Software development. **Prerequisite**: Basic understanding of Object Oriented Design and Development.

**Course Outcomes**: After the completion of the course the student will be able to

CO1	Differentiate Traditional and Agile Software Development approaches (Cognitive Knowledge Level: Understand)
CO2	Prepare Software Requirement <b>Specification</b> and Software Design for a given problem. ( <b>Cognitive Knowledge Level: Apply</b> )
CO3	Justify the significance of design patterns and licensing terms in software development, prepare testing, maintenance and DevOps strategies for a project. (Cognitive Knowledge Level: Apply)
CO4	Make use of software project management concepts while planning, estimation, scheduling, tracking and change management of a project, with proper application of SCRUM, Kanban and Lean frameworks. (Cognitive Knowledge Level: Apply)
CO5	Utilize SQA practices, Process Improvement techniques and Technology improvements <b>namely</b> cloud based software model and containers & microservices in a Software Development Process. (Cognitive Knowledge Level: <b>Apply</b> )

# Mapping of course outcomes with program outcomes

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO1 0	PO1 1	PO12
CO1	0	0	9	0		9						0
CO2	0	0	Ø	0		0				0	Ø	9

#### B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

CO3	0	0	0	0		0		0	0	0
CO4	0	0	0	0	0		0	0	0	0
CO5	0	0	0	0	0	KA	Ľ/	1V	1	0

	Abstract POs defined by National Board of Accreditation								
PO#	Broad PO	PO#	Broad PO						
PO1	Engineering Knowledge	PO7	Environment and Sustainability						
PO2	Problem Analysis	PO8	Ethics						
PO3	Design/Development of solutions	PO9	Individual and team work						
PO4	Conduct investigations of complex problems	PO10	Communication						
PO5	Modern tool usage	PO11	Project Management and Finance						
PO6	The Engineer and Society	PO12	Lifelong learning						

#### **Assessment Pattern**

Bloom's Category	Continuous As	End Semester	
	Test1 (Percentage)	Test2 (Percentage)	Examination Marks
Remember	30	30	30
Understand	30	30	30

#### B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

Apply	40	40	40
Analyse			
Evaluate	LARDI	II KAI	AM
Create	THNO	Incir	ΔI.

#### **Mark Distribution**

Total Marks	5	CIE Marks	ESE Marks	ESE Duration
150		50	100	3 hours

## **Continuous Internal Evaluation Pattern:**

Attendance		: 1 <mark>0</mark> marks
Continuous A	Assessment Tests	: 25 marks

Continuous Assessment Assignment : **15 marks** (Each student shall identify a software development problem and prepare Requirements Specification, Design Document, Project Plan and Test case documents for the identified problem as the assignment.)

## **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks.

First Internal Examination shall be preferably conducted after completing the first half of the syllabus and the Second Internal Examination shall be preferably conducted after completing the remaining part of the syllabus.

There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly covered module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly covered module), each with 7 marks. Out of the 7 questions in Part B, a student should answer any 5.

### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have a maximum of 2 subdivisions and carries 14 marks.

### **Syllabus**

## Module 1 : Introduction to Software Engineering (8 hours)

Introduction to Software Engineering - Professional software development, Software engineering ethics. Software process models - The waterfall model, Incremental development. Process activities - Software specification, Software design and implementation, Software validation, Software evolution. Coping with change - Prototyping, Incremental delivery, Boehm's Spiral Model. Agile software development - Agile methods, agile manifesto - values and principles. Agile development techniques, Agile Project Management. Case studies : An insulin pump control system. Mentcare - a patient information system for mental health care.

## Module 2 : Requirement Analysis and Design (10 hours)

Functional and non-functional requirements, Requirements engineering processes. Requirements elicitation, Requirements validation, Requirements change, Traceability Matrix. Developing use cases, Software Requirements Specification Template, Personas, Scenarios, User stories, Feature identification. Design concepts - Design within the context of software engineering, Design Process, Design concepts, Design Model. Architectural Design - Software Architecture, Architectural Styles, Architectural considerations, Architectural Design Component level design - What is a component?, Designing Class-Based Components, Conducting Component level design, Component level design for web-apps. Template of a Design Document as per "IEEE Std 1016-2009 IEEE Standard for Information Technology Systems Design Software Design Descriptions". Case study: The Ariane 5 launcher failure.

## Module 3 : Implementation and Testing (12 hours)

Object-oriented design using the UML, Design patterns, Implementation issues, Open-source development - Open-source licensing - GPL, LGPL, BSD. Review Techniques - Cost impact of Software Defects, Code review and statistical analysis. Informal Review, Formal Technical Reviews, Post-mortem evaluations. Software testing strategies - Unit Testing, Integration Testing, Validation testing, System testing, Debugging, White box testing, Path testing, Control Structure testing, Black box testing, Testing Documentation and Help facilities. Test automation, Test-driven development, Security testing. Overview of DevOps and Code Management - Code management, DevOps automation, CI/CD/CD. Software Evolution - Evolution processes, Software maintenance.

## Module 4 : Software Project Management (8 hours)

Software Project Management - Risk management, Managing people, Teamwork. Project Planning, Software pricing, Plan-driven development, Project scheduling, Agile planning. Estimation techniques, COCOMO cost modeling. Configuration management, Version management, System building, Change management, Release management, Agile software management - SCRUM framework. Kanban methodology and lean approaches.

## Module 5 : Software Quality and Process Improvement (6 hours)

Software Quality, Software Quality Dilemma, Achieving Software Quality Elements of Software Quality Assurance, SQA Tasks, Software measurement and metrics. Software Process Improvement(SPI), SPI Process CMMI process improvement framework, ISO 9001:2000 for Software.

# **Text Books**

- 1. Book 1 Ian Sommerville, Software Engineering, Pearson Education, Tenth edition, 2015.
- 2. Book 2 Roger S. Pressman, Software Engineering : A practitioner's approach, McGraw Hill publication, Eighth edition, 2014
- 3. Book 3 Ian Sommerville, Engineering Software Products: An Introduction to Modern Software Engineering, Pearson Education, First Edition, 2020.

## References

- 1. IEEE Std 830-1998 IEEE Recommended Practice for Software Requirements Specifications
- 2. IEEE Std 1016-2009 IEEE Standard for Information Technology—Systems Design— Software Design Descriptions
- 3. David J. Anderson, Kanban, Blue Hole Press 2010
- 4. David J. Anderson, Agile Management for Software Engineering, Pearson, 2003
- 5. Walker Royce, Software Project Management : A unified framework, Pearson Education, 1998
- 6. Steve. Denning, The age of agile, how smart companies are transforming the way work gets done. New York, Amacom, 2018.
- 7. Satya Nadella, Hit Refresh: The Quest to Rediscover Microsoft's Soul and Imagine a Better Future for Everyone, Harper Business, 2017
- 8. Henrico Dolfing, Project Failure Case Studies: Lessons learned from other people's mistakes, Kindle edition
- 9. Mary Poppendieck, Implementing Lean Software Development: From Concept to Cash, Addison-Wesley Signature Series, 2006
- 10. StarUML documentation https://docs.staruml.io/
- 11. OpenProject documentation https://docs.openproject.org/

- 12. BugZilla documentation https://www.bugzilla.org/docs/
- 13. GitHub documentation https://guides.github.com/
- 14. Jira documentation https://www.atlassian.com/software/jira

# **Course Level Assessment Questions**

# **Course Outcome 1 (CO1):**

- 1. What are the advantages of an incremental development model over a waterfall model?
- 2. Compare agile software development with traditional software development?

# Course Outcome 2 (CO2):

- 1. How to prepare a software requirement specification?
- 2. Differentiate between Architectural design and Component level design.
- 3. How do agile approaches help software developers to capture and define the user requirements effectively?
- 4. What is the relevance of the SRS specification in software development?
- 5. Prepare a use case diagram for a library management system.

# **Course Outcome 3 (CO3):**

- 1. Differentiate between the different types of software testing strategies.
- 2. What are the benefits of DevOps?
- 3. How do design patterns help software architects communicate the design of a complex system effectively?
- 4. What are the proactive approaches one can take to optimise efforts in the testing phase?

# **Course Outcome 4 (CO4):**

- 1. What are the activities involved in software project management?
- 2. What is the need for SCRUM, Kanban and Lean methodologies?
- 3. What are the benefits of rolling level planning in software project management and how would you implement it?
- 4. How would you assess the risks in your software development project? How would you plan for risk mitigation and contingency?

# **Course Outcome 5 (CO5):**

- 1. What is the importance of Software Process improvement?
- 2. How will retrospectives help in improving the software development process?
- 3. What are the important skills required for the SQA role?
- 4. How would you use project history data as a prediction tool to plan future projects?

# B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

# **Model Question Paper**

	QP CODE:
	Reg No:
	Name :
	PAGES : 3 APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY
F	IFTH SEMESTER B.TECH DEGREE EXAMINATION(MINOR), MONTH & YEAR
•	Course Code: CST 381
	Course Name: Concepts in Software Engineering
	Duration: 3 Hrs Max. Marks : 100
	PART A
	Answer all Questions. Each question carries 3 Marks
1.	Explain why professional software that is developed for a customer is not simply
	the programs that have been developed and delivered
2.	Incremental software development could be very effectively used for customers
2.	who do not have a clear idea about the systems needed for their operations.
	Discuss.
3.	Identify and briefly describe four types of requirements that may be defined for a
	computer based system.
4.	Describe software architecture in your own words.
4.	Describe software areintecture in your own words.
5	What are the major differences between CDL and LCDL?
5.	What are the major differences between GPL and LGPL?
6.	Compare between white box testing and black box testing.
	2014
7.	What is the importance of risk management in software project management?
8.	Explain COCOMO cost estimation model
9.	Describe the software quality dilemma in your own words
10.	Which are the levels of the CMMI model?
	(10x3=3

**30**)

# Part B (Answer any one question from each module. Each question carries 14 marks)

<b>11.</b> (a	) Compare between waterfall model and spiral model	(8)
(b	) Explain Agile methods and Agile manifesto OR	(6)
<b>12.</b> (a	) Explain software process activities	(7)
(b	) Explain Agile Development techniques and Agile Project Management.	(7)
13. (a	) What are functional and nonfunctional requirements? Imagine that you are developing a library management software for your college, identify at least 8 functional requirements and 4 nonfunctional requirements.	(10)
(b	) What are the contents of a software requirement specification?	(4)
<b>14.</b> (a	) Explain Personas, Scenarios, User stories and Feature identification?	(8)
(b	) Compare between Software Architecture design and Component level design	(6)
<b>15.</b> (a	) Describe the formal and informal review techniques in detail.	(6)
(b	) Explain various software testing strategies.	(8)
	OR 2014	
16. (a	) Explain DevOps CI/CD/CD in detail.	(8)
(b	) Explain test driven development.	(6)
17. (a	) What is a critical path and demonstrate its significance in a project schedule with the help of a sample project schedule.	(6)
(b	) Explain plan driven development and project scheduling	(6)

# OR

18.	(a)	Explain the SCRUM framework.	(8)
	(b)	What is algorithmic cost modeling? What problems does it suffer from when compared with other approaches to cost estimation?	(6)
19.	(a)	Explain elements of Software Quality Assurance and SQA Tasks.	(8)
	(b)	Explain the SPI process.	(6)
		OR	
20.	(a)	Compare between CMMI and ISO 9001:2000	(8)
	(b)	Compare Quality Control and Quality Assurance.	(6)

	Teaching Plan [44 hours]	
	Module 1 : Introduction to Software Engineering (8 hours)	Hours
1.1	Introduction to Software Engineering. [Book 1, Chapter 1]	1 hour
1.2	Software process models [Book 1 - Chapter 2]	1 hour
1.3	Process activities [Book 1 - Chapter 2]	1 hour
1.4	Coping with change [Book 1 - Chapter 2, Book 2 - Chapter 4]	1 hour
1.5	Agile software development [Book 1 - Chapter 3]	1 hour
1.6	Agile development techniques [Book 1 - Chapter 3]	1 hour
1.7	Agile Project Management.[Book 1 - Chapter 3]	1 hour
1.8	Case studies : An insulin pump control system. Mentcare - a patient information system for mental health care. [Book 1 - Chapter 1]	1 hour
	Module 2 : Requirement Analysis and Design (10 hours)	
2.1	Functional and non-functional requirements, Requirements engineering processes [Book 1 - Chapter 4]	1 hour

2.2	Requirements elicitation, Requirements validation, Requirements change, Traceability Matrix [Book 1 - Chapter 4]			
2.3	Developing use cases, Software Requirements Specification Template [Book 2 - Chapter 8]	1 hour		
2.4	Personas, Scenarios [Book 3 - Chapter 3]	1 hour		
2.5	User stories, Feature identification [Book 3 - Chapter 3]	1 hour		
2.6	Design concepts [Book 2 - Chapter 12]	1 hour		
2.7	Architectural Design [Book 2 - Chapter 13]	1 hour		
2.8	Component level design [Book 2 - Chapter 14]	1 hour		
2.9	Component level design, Design Document Template. [Book 2 - Chapter 14, Ref - 2]	1 hour		
2.10	Case study: The Ariane 5 launcher failure. [Book 2 - Chapter 16]	1 hour		
	Module 3 : Implementation and Testing (12 hours)	1		
3.1	Object-oriented design using the UML, Design patterns [Book 1 - Chapter 7]	1 hour		
3.2	Implementation issues, Open-source development - Open-source licensing - GPL, LGPL, BSD [Book 1 - Chapter 7]	1 hour		
3.3	Review Techniques - Cost impact of Software Defects, Code review. [Book 2 - Chapter 20]	1 hour		
34	Informal Review, Formal Technical Reviews, Post-mortem evaluations. [Book 2 - Chapter 20]	1 hour		
3.5	Software testing strategies [Book 2 - Chapter 22]	1 hour		
3.6	Software testing strategies [Book 2 - Chapter 22]	1 hour		
3.7	White box testing, Path testing, Control Structure testing [Book 2 - Chapter 23]	1 hour		
3.8	Black box testing. Test documentation [Book 2 - Chapter 23]	1 hour		
3.9	Test automation, Test-driven development [Book 3 - Chapter 9]	1 hour		
3.10	Security testing. DevOps and Code Management [Book 3 - Chapter 9, Chapter 10]	1 hour		
3.11	DevOps and Code Management - Code management, DevOps automation, CI/CD/CD. [Book 3 - Chapter 10]	1 hour		

3.12	Software Evolution - Evolution processes, Software maintenance. [Book 1 - Chapter 9]	1 hour
	Module 4 : Software Project Management (8 hours)	
4.1	Software Project Management - Risk management, Managing people, Teamwork [Book 1 - Chapter 22]	1 hour
4.2	Project Planning - Software pricing, Plan-driven development, Project scheduling, Agile planning [Book 1 - Chapter 23]	1 hour
4.3	Estimation techniques [Book 1 - Chapter 23]	1 hour
4.4	Configuration management [Book 1 - Chapter 25]	1 hour
4.5	Agile software management - SCRUM framework [Book 2 - Chapter 5]	1 hour
4.6	Agile software management - SCRUM framework [Book 2 - Chapter 5]	1 hour
4.7	Kanban methodology and lean approaches. [Ref 9 - Chapter 2]	1 hour
4.8	Kanban methodology and lean approaches.[Ref 9 - Chapter 2]	1 hour
Modu	le 5 : Software Quality, Process Improvement and Technology trends (6 hour	s)
5.1	Software Quality, Software Quality Dilemma, Achieving Software Quality. [Book 2 - Chapter 19]	1 hour
5.2	Elements of Software Quality Assurance, SQA Tasks [Book 3 - Chapter 21]	1 hour
5.3	Software measurement and metrics. [Book 3 - Chapter 21]	1 hour
5.4	Software Process Improvement(SPI), SPI Process[Book 2 - Chapter 37]	1 hour
5.5	Software Process Improvement(SPI), SPI Process[Book 2 - Chapter 37]	1 hour
5.6	CMMI process improvement framework, ISO 9001:2000 for Software. [Book 2 - Chapter 37]	1 hour

CST 383	CONCEPTS IN MACHINE LEARNING	Category	L	Т	Р	Credit	Year of introduction
		VAC	3	1	0	4	2019

**Preamble**: This course enables the learners to understand the fundamental concepts and algorithms in machine learning. The course covers the standard and most popular supervised learning algorithms such as linear regression, logistic regression, decision trees, Bayesian learning & the naive Bayes algorithm, support vector machines& kernels, basic clustering algorithms and dimensionality reduction methods. This course helps the students to provide machine learning based solutions to real world problems.

**Prerequisite:** Familiarity with basics in linear algebra, probability and Python

programming.

	Course Outcomes					
CO1	Illustrate Machine Learning concepts and basic parameter estimation methods.(Cognitive Knowledge Level: Apply)					
CO2	Demonstrate supervised learning concepts (regression, linear classification). (Cognitive Knowledge Level: Apply)					
CO3	Illustrate the concepts of Multilayer neural network and Support Vector Machine (Cognitive Knowledge Level: Apply)					
CO4	Describe unsupervised learning concepts and dimensionality reduction techniques. (Cognitive Knowledge Level: Apply)					
CO5	Solve real life problems using appropriate machine learning models and evaluate the performance measures ( <b>Cognitive Knowledge Level: Apply</b> )					

# Mapping of course outcomes with program outcomes

	PO1	PO2	PO3	PO4	PO5	PO6	<b>PO</b> 7	PO8	PO9	PO10	PO11	PO1 2
CO1	$\bigcirc$	$\bigcirc$	$\bigcirc$	$\bigcirc$	$\bigcirc$							$\bigcirc$
CO2	$\bigcirc$	$\bigcirc$	$\oslash$	$\oslash$	$\oslash$							$\oslash$
CO3	$\bigcirc$	$\bigcirc$	$\oslash$	$\oslash$	$\oslash$							$\oslash$

### B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

CO4	$\bigcirc$	$\bigcirc$	$\bigcirc$	$\bigcirc$	$\bigcirc$				$\oslash$
CO5	$\bigcirc$	$\bigcirc$	$\bigcirc$	$\bigcirc$	$\bigcirc$	$\bigcirc$			$\bigcirc$

	Abstract POs defined by National Board of Accreditation								
PO#	Broad PO	PO#	Broad PO						
PO1	Engineering Knowledge	PO7	Environment and Sustainability						
PO2	Problem Analysis	PO8	Ethics						
PO3	Design/Development of solutions	PO9	Individual and team work						
PO4	Conduct investigations of complex problems	PO10	Communication						
PO5	Modern tool usage	PO11	Project Management and Finance						
PO6	The Engineer and Society	PO12	Life long learning						

## **Assessment Pattern**

Bloom's Category	Continuous Assessme	End Semester		
	Test1 (Percentage)	Test2 (Percentage)	Examination Marks	
Remember	30	30	30	
Understand	30	30	30	
Apply	40	40	40	
Analyse				
Evaluate				
Create		124		

# **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3 hours

# **Continuous Internal Evaluation Pattern:**

Attendance	: 10 marks
Continuous Assessment Tests	: 25 marks

# Continuous Assessment Assignment : 15 marks

# **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks

First Internal Examination shall be preferably conducted after completing the first half of the syllabus and the Second Internal Examination shall be preferably conducted after completing remaining part of the syllabus.

There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly covered module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly covered module), each with 7 marks. Out of the 7 questions in Part B, a student should answer any 5.

# **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have maximum 2 sub-divisions and carry 14 marks.

## Syllabus

# Module-1 (Overview of machine learning)

Machine learning paradigms-supervised, semi-supervised, unsupervised, reinforcement learning. Basics of parameter estimation - maximum likelihood estimation(MLE) and maximum a posteriori estimation(MAP). Introduction to Bayesian formulation.

## Module-2 (Supervised Learning)

Regression - Linear regression with one variable, Linear regression with multiple variables, solution using gradient descent algorithm and matrix method, basic idea of overfitting in regression. Linear Methods for Classification- Logistic regression, Perceptron, Naive Bayes, Decision tree algorithm ID3.

# Module-3 (Neural Networks (NN) and Support Vector Machines (SVM))

NN - Multilayer feed forward network, Activation functions (Sigmoid, ReLU, Tanh), Backpropagation algorithm.

SVM - Introduction, Maximum Margin Classification, Mathematics behind Maximum Margin Classification, Maximum Margin linear separators, soft margin SVM classifier, non-linear SVM, Kernels for learning non-linear functions, polynomial kernel, Radial Basis Function(RBF).

## Module-4 (Unsupervised Learning)

Clustering - Similarity measures, Hierarchical Agglomerative Clustering, K-means partitional clustering, Expectation maximization (EM) for soft clustering. Dimensionality reduction – Principal Component Analysis, factor Analysis, Multidimensional scaling, Linear Discriminant Analysis.

## Module-5 (Classification Assessment)

Classification Performance measures - Precision, Recall, Accuracy, F-Measure, Receiver Operating Characteristic Curve(ROC), Area Under Curve(AUC. Bootstrapping, Cross Validation, Ensemble methods, Bias-Variance decomposition. Case Study: Develop a classifier for face detection.

# **Text Book**

- 1. Ethem Alpaydin, Introduction to Machine Learning, 2nd edition, MIT Press 2010.
- Mohammed J. Zaki and Wagner Meira, Data Mining and Analysis: Fundamental Concepts and Algorithms, Cambridge University Press, First South Asia edition, 2016.
- 3. Jake VanderPlas, Python Data Science Handbook, O'Reilly Media, 2016
- 4. Tom Mitchell, Machine Learning, McGraw-Hill, 1997.

# **Reference Books**

- 1. Christopher Bishop. Neural Networks for Pattern Recognition, Oxford University Press, 1995.
- 2. Kevin P. Murphy. Machine Learning: A Probabilistic Perspective, MIT Press 2012.
- 3. Trevor Hastie, Robert Tibshirani, Jerome Friedman, The Elements Of Statistical Learning, Second edition Springer 2007.
- 4. P. Langley, Elements of Machine Learning, Morgan Kaufmann, 1995.
- 5. Richert and Coelho, Building Machine Learning Systems with Python.
- Davy Cielen, Arno DB Meysman and Mohamed Ali. Introducing Data Science: Big Data, Machine Learning, and More, Using Python Tools, Dreamtech Press 2016.

### Sample Course Level Assessment Questions

## **Course Outcome1(CO1):**

- 1. A coin is tossed 100 times and lands heads 62 times. What is the maximum likelihood estimate for  $\theta$ , the probability of heads.
- 2. Suppose data  $x_1, ..., x_n$  are independent and identically distributed drawn from an exponential distribution  $exp(\lambda)$ . Find the maximum likelihood for  $\lambda$ .
- 3. Suppose  $x_1, ..., x_n$  are independent and identically distributed(iid) samples from a distribution with density

$$f_X(x \mid \theta) = \begin{cases} \frac{\theta x^{\theta - 1}}{3^{\theta}}, & 0 \le x \le 3\\ 0, & \text{otherwise} \end{cases}$$

Find the maximum likelihood estimate(MLE) for  $\theta$ .

4. Find the maximum likelihood estimator (MLE) and maximum a posteriori (MAP) estimator for the mean of a univariate normal distribution. Assume that we have N samples,  $x_1, ..., x_N$  independently drawn from a normal distribution with known variance  $\sigma^2$  and unknown mean  $\mu$  and the prior distribution for the mean is itself a normal distribution with mean v and variance  $\beta^2$ . What happens to the MLE and MAP estimators as the number of samples goes to infinity.

### **Course Outcome 2 (CO2):**

- 1. Explain the difference between (batch) gradient descent and stochastic gradient descent. Give an example of when you might prefer one over the other.
- 2. Suppose that you are asked to perform linear regression to learn the function that outputs y, given the *D*-dimensional input x. You are given N independent data points, and that all the D attributes are linearly independent. Assuming that D is around 100, would you prefer the closed form solution or gradient descent to estimate the regressor?
- 3. Suppose you have a three class problem where class label  $y \in 0$ , 1, 2 and each training example X has 3 binary attributes  $X_1, X_2, X_3 \in 0$ , 1. How many parameters (probability distribution) do you need to know to classify an example using the Naive Bayes classifier?

## Course Outcome 3 (CO3):

- 1. What are support vectors and list any three properties of the support vector classifier solution?
- 2. Why do you use kernels to model a projection from attributes into a feature space, instead of simply projecting the dataset directly?
- 3. Describe how Support Vector Machines can be extended to make use of kernels. Illustrate with reference to the Gaussian kernel  $K(x, y) = e^{-z}$ , where  $z = (x-y)^2$ .

- 4. Briefly explain one way in which using tanh instead of logistic activations makes optimization easier.
- 5. ReLU activation functions are most used in neural networks instead of the tanh activation function. Draw both activation functions and give a) an advantage of the ReLU function compared to the tanh function. b) a disadvantage of the ReLU function compared to the tanh function.

## **Course Outcome 4(CO4):**

- 1. Describe cluster analysis? Identify two applications where cluster analysis can be applied to multimedia data?
- 2. Given two objects represented by the tuples (22, 1, 42, 10) and (20, 0, 36, 8):
- (i) Compute the Euclidean distance between the two objects.
- (ii) Compute the Manhattan distance between the two objects.
- 3. Use PCA to reduce the dimension from 2 to 1 for the design matrix *X*.

$$X = \begin{bmatrix} 6 & -4 \\ -3 & 5 \\ -2 & 6 \\ 7 & -3 \end{bmatrix}$$

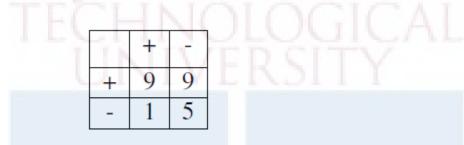
- 4. What is Principal Component Analysis (PCA)? Which eigen value indicates the direction of largest variance?
- 5. Suppose that one runs a principal component analysis on a data set and tells that the percentage of variance explained by the first 3 components is 80%. How is this percentage of variance explained?

# Course Outcome 5 (CO5):

- 1. Suppose that you are contacted by a food processing company that wants you to develop a classifier that detects whether a rat is present in an image. You collect a large dataset of images by crawling the web, and have annotators determine which images contain rats. This set of images can then be used as the training set for your classifier.
  - a. Suggest a machine learning method to use for this classification task and evaluate its performance.
  - b. After you have delivered your solution to the company, they get back to you and complain that when they evaluate on a new test set, they get precision and recall values that are much lower than what you reported to them. Explain what might have gone wrong and propose remedial measures .
- 2. A real estate firm would like to build a system that predicts the sale prices of a house. They create a spreadsheet containing information about 1,500 house sales in the Kochi

area. In addition to the price, there are 10 features describing the house, such as number of bedrooms, total indoor area, lot area, a swimming pool, location, etc. Explain how you would implement a machine learning model that would solve this prediction task. Give all steps you would carry out when developing it. Explain why the model you built is probably useless in the long run.

3. For a classifier, the confusion matrix is given by:



What is the precision, recall and accuracy of that classifier?

**Model Question Paper** 

PAGES:3

**QP CODE:** 

Reg No:\_ Name:

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY FIFTH SEMESTER B.TECH DEGREE EXAMINATION(MINOR), MONTH & YEAR

# Course Code: CST 383

# **Course Name: CONCEPTS IN MACHINE LEARNING**

Max.Marks:100 Hours

### **Duration: 3**

# PART A

## Answer all Questions. Each question carries 3 Marks

- 1. Define supervised learning? Name special cases of supervised learning depending on whether the inputs/outputs are categorical, or continuous.
- 2. Differentiate between Maximum Likelihood estimation (MLE) and Maximum a Posteriori (MAP) estimation?
- 3. What is overfitting and why is it a problem? Give an example of a method to reduce the risk of overfitting.
- 4. Specify the basic principle of gradient descent algorithm.
- 5. Suppose that you have a linear support vector machine(SVM) binary classifier. Consider a point that is currently classified correctly, and is far away from the decision boundary. If you

remove the point from the training set, and re-train the classifier, will the decision boundary change or stay the same? Justify your answer.

- 6. Mention the primary motivation for using the kernel trick in machine learning algorithms?
- 7. Expectation maximization (EM) is designed to find a maximum likelihood setting of the parameters of a model when some of the data is missing. Does the algorithm converge? If so, do you obtain a locally or globally optimal set of parameters?
- 8. Illustrate the strength and weakness of k-means algorithm.
- 9. Classifier A attains 100% accuracy on the training set and 70% accuracy on the test set. Classifier B attains 70% accuracy on the training set and 75% accuracy on the test set. Which one is a better classifier. Justify your answer.
- 10. How does bias and variance trade-off affect machine learning algorithms?

(10x3=30)

### Part B

### Answer any one Question from each module. Each question carries 14 Marks

11. a) Suppose that X is a discrete random variable with the following probability mass function: where  $\theta \le \theta \le 1$  is a parameter. The following 10 independent observations

X	0	1	2	3
P(X)	$2\theta/3$	$\theta/3$	$2(1-\theta)/3$	$(1 - \theta)/3$

were taken from such a distribution: (3, 0, 2, 1, 3, 2, 1, 0, 2, 1). What is the maximum likelihood estimate of  $\theta$ . (6)

b) A gamma distribution with parameters  $\alpha$ ,  $\beta$  has the following density function, where  $\Gamma(t)$  is the gamma function.

$$p(x) = \frac{\beta^{\alpha}}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x}$$

If the posterior distribution is in the same family as the prior distribution, then we say that the prior distribution is the conjugate prior for the likelihood function. Using the Gamma distribution as a prior, show that the Exponential distribution is a conjugate prior of the Gamma distribution. Also, find the maximum a posteriori estimator for the parameter of the Exponential distribution as a function of  $\alpha$  and  $\beta$ . (8)

OR

12. a) Traffic between 8AM and 9AM at a certain place was measured by counting the number of vehicles that passed at that time. Suppose the counts follow a Poisson process. A random sample of 9 observations was collected, having observed the following number of vehicles: (95, 100, 80, 70, 110, 98, 97, 90, 70). Derive the maximum likelihood estimator for the

(4)

average number of vehicles that pass by that place between 8 AM and 9 AM, and compute the corresponding estimate using the given sample. (7)

- b) Find the maximum a posteriori (MAP) estimator for the mean of a univariate normal distribution. Assume that we have N samples,  $x_1, ..., x_N$  independently drawn from a normal distribution with known variance  $\sigma^2$  and unknown mean  $\mu$  and the prior distribution for the mean is itself a normal distribution with mean v and variance  $\beta^2$ . (7)
- 13.a) Derive the gradient descent training rule assuming for the target function  $o_d = w_0 + w_1 x_1 + ...$ +  $w_n x_n$ . Define explicitly the squared cost/error function E, assuming that a set of training examples D is provided, where each training example  $d \in D$  is associated with the target output  $t_d$ . (10)
- b) How can we interpret the output of a two-class logistic regression classifier as a probability?

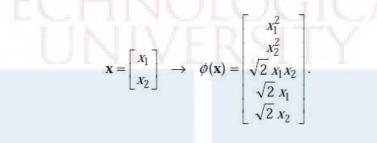
#### OR

- 14. a) In a two-class logistic regression model, the weight vector w = [4, 3, 2, 1, 0]. We apply it to some object that we would like to classify; the vectorized feature representation of this object is x = [-2, 0, -3, 0.5, 3]. What is the probability, according to the model, that this instance belongs to the positive class? (6)
- b) The following dataset can be used to train a classifier that determines whether a given person is likely to own a car or not. There are three features: education level (primary, secondary, or university); residence (city or country); gender (female, male).

education	residence	gender	has car?
sec	country	female	yes
univ	country	female	yes
prim	city	male	no
univ	city	male	no
sec	city	female	no
sec	country	male	yes
prim	country	female	yes
univ	country	male	yes
sec	city	male	yes
prim	city	female	no
univ	city	female	no
prim	country	male	yes

Find the root attribute and justify your answer

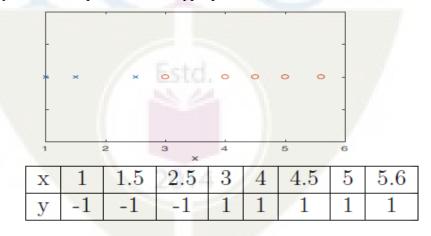
15. a) Consider a support vector machine whose input space is 2-D, and the inner products are computed by means of the kernel  $K(x, y) = (x \cdot y + 1)^2 \cdot 1$ , where x.ydenotes the ordinary inner product. Show that the mapping to feature space that is implicitly defined by this kernel is the mapping to 5-D given by (10)



b) What is the basic idea of a Support Vector Machine?

OR

- 16. a) Explain how back propagation can be used to solve XOR problem which is not linearly separable.(8)
- b) Consider the following one dimensional training data set, 'x' denotes negative examples and 'o' positive examples. The exact data points and their labels are given in the table. Suppose a SVM is used to classify this data. Indicate which are the support vectors and mark the decision boundary. Find the equation of the hyperplane. (6)



- 17. a)Suppose that we have the following data (one variable). Use single linkage Agglomerative clustering to identify the clusters.Daa: (2, 5, 9, 15, 16, 18, 25, 33, 33, 45).
- b) Given two objects represented by the tuples (22, 1, 42, 10) and (20, 0, 36, 8):
- (i) Compute the Euclidean distance between the two objects.
- (ii) Compute the Manhattan distance between the two objects.

(8)

(4)

(8)

(6)

(iii) Compute the Minkowski distance between the two objects, using p = 3

OR

 18. a) Suppose that we have the following data:

 
$$\overline{a \ b \ c \ d \ e \ f \ g \ h \ i \ j}}$$
 $\overline{a \ (2,0) \ (1,2) \ (2,2) \ (3,2) \ (2,3) \ (3,3) \ (2,4) \ (3,4) \ (4,4) \ (3,5)}}$ 

Identify the cluster by applying the k-means algorithm, with k = 2. Try using initial cluster centers as far apart as possible. (10)(4)

b) List the steps involved in Principal Component Analysis.

**19.** a) Suppose the dataset had 9700 cancer-free images from 10000 images from cancer patients. Find precision, recall and accuracy ? Is it a good classifier? Justify. (8)

Actual Class\ Predicted class	cancer = yes	cancer = no	Total
cancer = yes	90	210	300
cancer = no	140	9560	9700
Total	230	9770	10000

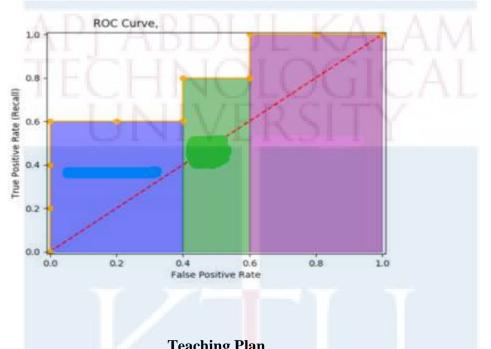
b) Suppose that you have a classification problem where our feature representation contains about 10,000,000 features. We would like to develop a classifier that can be deployed in a mobile phone, so preferably it should have a small memory footprint. Discuss one solution for how this can be done. (6)

#### OR

- 20. a) What are ROC space and ROC curve in machine learning? In ROC space, which points correspond to perfect prediction, always positive prediction and always negative prediction? Why? (6)
- b)Suppose there are three classifiers A,B and C. The (FPR, TPR) measures of the three classifiers are as follows – A (0, 1), B (1, 1), C (1,0.5). Which can be considered as a perfect classifier? Justify your answer. (4)

# c) Given the following ROC Curve? Find the AUC?

(4)



	raciiii	ig i ia	11	
100				

No	Contents	No of Lecture Hrs
	Module 1 :Overview of machine learning (7 hours)	
1.1	Supervised, semi-supervised, unsupervised learning, reinforcement learning (Text Book (TB) 1: Chapter 1)	1hour
1.2	Maximum likelihood estimation(MLE) (TB 1: Section 4.2)	1hour
1.3	Maximum likelihood estimation (MLE)- example (TB 1: Section 4.2)	1hour
1.4	Maximum a posteriori estimation(MAP) (TB 4: Section 6.2)	1hour
1.5	Maximum a posteriori estimation(MAP)-example (TB 4: Section 6.2)	1hour
1.6	Bayesian formulation (TB 1: Section 14.1, 14.2)	1hour
1.7	Bayesian formulation -example (TB 1: Section 14.1, 14.2)	1hour
	Module 2 : Supervised Learning (8 hours)	

2.1	Linear regression with one variable (TB 1: Section 2.6)	1hour				
2.2	Multiple variables, Solution using gradient descent algorithm and matrix method (No derivation required) (TB 1: Section 5.8)	1hour				
2.3	Overfitting in regression, Lasso and Ridge regularization	1hour				
2.4	Logistic regression					
2.5	Perceptron	1hour				
2.6	Naive Bayes (TB 2: Section 18.2)	1hour				
2.7	Decision trees (TB 2: Chapter 19)	1hour				
2.8	Decision trees- ID3 algorithm (TB 2: Chapter 19)	1hour				
Modu	Ile 3 : Neural Networks and Support Vector Machines (TB 2: Chapter 21)					
	(11 hours)					
3.1	Multilayer Feed forward Network, Activation Functions (Sigmoid, ReLU, Tanh)	1hour				
3.2	Back Propagation Algorithm					
3.3	Illustrative Example for Back Propagation					
3.4	Introduction, Maximum Margin Hyperplane,	1hour				
3.5	Mathematics behind Maximum Margin Classification	1hour				
3.6	Formulation of maximum margin hyperplane and solution	1hour				
3.7	Soft margin SVM	1hour				
3.8	Solution of Soft margin SVM	1hour				
3.9	Non-linear SVM	1hour				
3.10	Kernels for learning non-linear functions and properties of kernel functions.	1hour				
3.11	Example Kernels functions- Linear, RBF, Polynomial.	1hour				
	Module 4 : Unsupervised Learning (10 hours)					
4.1	Similarity measures- Minkowski distance measures( Manhattan, Euclidean), Cosine Similarity	1hour				
4.2	Clustering - Hierarchical Clustering (TB 2: Chapter 14)	1hour				
4.3	K-means partitional clustering (TB 2: Chapter 13)	1hour				
4.4	Expectation maximization (EM) for soft clustering (TB 2: Chapter 13)	1hour				
4.5	Expectation maximization (EM) for soft clustering (TB 2: Chapter 13)	1hour				

4.6	Dimensionality reduction – Principal Component Analysis (TB 1: Section 6.3)	1hour
4.7	Dimensionality reduction – Principal Component Analysis (TB 1: Section 6.3)	1hour
4.8	Factor Analysis (TB 1: Section 6.4)	1hour
4.9	Multidimensional scaling (TB 1: Section 6.5)	1hour
4.10	Linear Discriminant Analysis (TB 1: Section 6.6)	1hour
	Module 5 : Classification Assessment (8 hours)	
5.1	Performance measures - Precision, Recall, Accuracy, F-Measure, ROC, AUC. (TB 2: Chapter 22.1)	1hour
5.2	Boot strapping, Cross validation	1hour
5.3	Ensemble methods- bagging	1hour
5.4	Ensemble methods- boosting	1hour
5.5	Bias-Variance decomposition (TB 2: Chapter 22.3)	1hour
5.6	Bias-Variance decomposition (TB 2: Chapter 22.3)	1hour
5.7	Face detection (TB 3: Chapter 5 Section Application: A Face Detection Pipeline)	1hour
5.8	Face detection (TB 3: Chapter 5 Section Application: A Face Detection Pipeline)	1hour



CST 385	CLIENT SERVER SYSTEMS	Category	L	Т	Р	Credit	Year of Introduction
505	ADI	VAC	3	1	0		2019

# **Preamble:**

The syllabus is prepared with the view of preparing the Engineering Graduates to build effective Client/Server applications. This course aims at providing a foundation in decentralized computer systems, using the client/server model. The course content is decided to cover the essential fundamentals which can be taught within the given slots in the curriculum.

# Prerequisite: Basic knowledge in Computer

Course Outcomes: After the completion of the course the student will be able to

	Course Outcomes
	Course Outcomes
CO 1	Identify the basics of client/server systems and the driving force behind the
001	development of client/server systems(Cognitive Knowledge Level: Understand)
CO 2	Outline the architecture and classifications of client/server systems(Cognitive
	Knowledge Level: Understand)
	Estd.
CO 3	Summarize the client/server network services for an application(Cognitive
05	Knowledge Level: Understand)
CO 4	Identify management services and issues in network (Cognitive Knowledge Level:
004	Understand)
	2017
CO 5	Outline the Client/Server technology in respect of databases and Client/Server
0.05	database architecture (Cognitive Knowledge Level: Understand)

	PO1	PO2	PO3	PO4	PO5	PO6	<b>PO7</b>	PO8	PO9	<b>PO10</b>	PO11	PO12
CO1	0	0	A	B			K	A	A	M		Ø
CO2		0	ŪÌ	Ϋ́Ι	VI	ÉR	SÌ	Ť	Y	1.1.		0
CO3	0	Ø										$\oslash$
CO4	0											$\oslash$
CO5	0	$\oslash$										$\oslash$

# Mapping of course outcomes with program outcomes

	Abstract POs defined by Natio	nal Board	l of Accreditation
PO#	Broad PO	PO#	Broad PO
PO1	Engineering Knowledge	PO7	Environment and Sustainability
PO2	Problem Analysis	PO8	Ethics
PO3	Design/Development of solutions	PO9	Individual and team work
PO4	Conduct investigations of complex problems	PO10	Communication
PO5	Modern tool usage	PO11	Project Management and Finance
PO6	The Engineer and Society	PO12	Life long learning

Bloom's	Continuous Assessme	End Semester		
Category			Examination Marks	
	Test 1 (Percentage)	Test2(Percentage)		
Remember	40	40	40	
Understand	40 V	40	40	
Apply	20	20	20	
Analyse				
Evaluate				
Create				

# **Assessment Pattern**

# Mark distribution

Total	CIE	ESE	ESE Duration
Marks	Marks	Marks	
150	50	100	3 hours

# **Continuous Internal Evaluation Pattern:**

Attendance	: 10 marks
Continuous Assessment Test 1 (for theory, for 2 hrs)	: 20 marks

Continuous Assessment Test 2 (for lab, internal examination, for 2hrs) : 20 marks

# **Internal Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 5 questions with 2 questions from each module (2.5 modules x = 5), having 3 marks for each question. Students should answer all questions. Part B also contains 5 questions with 2 questions from each module (2.5 modules

x = 5, of which a student should answer any one. The questions should not have subdivisions and each one carries 7 marks.

### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have maximum 2 sub-divisions and carry 14 marks.

## Sample Course Level Assessment Questions

### **Course Outcome 1 (CO1):**

1. Computing in client server architecture over Mainframe architecture has certain advantages and disadvantages. Describe at least three advantages and disadvantages for each architecture.

### Course Outcome 2 (CO2):

**1.** Explain the role of mainframe-centric model in Client/Server computing?

## Course Outcome 3(CO3):

1. Describe the client server system development methodology? Explain different phases of System Integration Life-Cycle.

### **Course Outcome 4 (CO4):**

**1.** Explain about network management and remote system management. How can security be provided to the network?

### **Course Outcome 5 (CO5):**

1. Explain various types of Client/Server Database Architecture

#### **Syllabus**

### Module – 1 (Introduction)

Introduction to Client/Server computing - Basic Client/Server Computing Model, Server for Every Client- File Server, Print Server, Application Server, Mail Server, Directory Services Server, Web Server, Database Server, Transaction Servers. Client/Server-Fat or Thin, Stateless

or Stateful, Servers and Mainframes, Client/Server Functions. Driving Forces behind Client/Server Computing- Business Perspective, Technology Perspective.

## Module -2 (Client/Server Classification)

Client/Server Types-Single Client/Single Server, Multiple Clients/Single Server, Multiple Clients/Multiple Servers, Integration With Distributed Computing, Alternatives To Client/Server Systems. Classification of Client/Server Systems- Two-Tier Computing, Middleware, Three-Tier Computing- Model View Controller (MVC), Principles behind Client/Server Systems.Client/Server Topologies. Existing Client/Server Architecture. Architecture for Business Information System.

# Module -3 (Client/Server Application Components)

Client- Services, Request for services, RPC, Windows services, Print services, Remote boot services, other remote services, Utility Services. Server- Detailed server functionality, Network operating system, Available platforms, Server operating system. Organizational Expectations, Improving performance of client/server applications, Single system image, Downsizing and Rightsizing, Advantages and disadvantages of Client/Server computing, Applications of Client/Server.

## Module -4 (Client/ Server Systems Services and Support)

Services and Support- System administration, Availability, Reliability, Scalability, Observability, Agility, Serviceability. Software Distribution, Performance, Network management. Remote Systems Management- RDP, Telnet, SSH, Security. LAN and Network Management issues.

## Module -5(Client/Server Technology and Databases)

Client/Server Technology and Databases - Storing Data, Database System Architectures. Client/Server In Respect Of Databases- Client/Server Databases, Client/Server Database Computing, Database Computing Vs. Mainframe, PC/File Server Computing. Client/Server Database Architecture - Process-Per-Client Architecture, Multi-Threaded Architecture, Hybrid Architecture. Database Middleware Component - Application Programming Interface, Database Translator, Network Translator.

## **Text Book**

- 1. Patrick Smith & Steve Guengerich, Client / Server Computing, PHI
- 2. Subhash Chandra Yadav, Sanjay Kumar Singh, An Introduction to Client/Server Computing, New Age International Publishers

## **Reference Books**

- 1. Jeffrey D.Schank, "Novell's Guide to Client-Server Application & Architecture" Novell Press
- 2. Robert Orfali, Dan Harkey, Jeri Edwards, Client/Server Survival Guide, Wiley-India Edition, Third Edition
- 3. Dawna Travis Dewire, Client Server Computing McGraw Hill
- 4. W.H.Inman, Developing Client Server Applications, BPB

## **Model Question Paper**

**QP CODE:** 

Reg No:\_\_\_\_\_

Name:

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

## FIFTH SEMESTER B.TECH DEGREE EXAMINATION(MINOR), MONTH & YEAR

## Course Code: CST 385

## Course Name : Client Server Systems

Max Marks: 100

**Duration: 3 Hours** 

PAGES:

## PART-A

## (Answer All Questions. Each question carries 3 marks)

- 1. Differentiate between Stateful and Stateless servers
- 2. List the different phases and activities of client/server system development methodology.
- 3. How does transmission protocol work in client/server applications?
- 4. List any six services in single system image environment.
- 5. Specify the role of the client in Client/Server computing and also list any six services provided by the client.
- 6. Why do most RPC system support call by value semantics for parameter passing?
- 7. What do you mean by a thin client network? List three advantages of the Thin

(10x3=30)

Client Network system.

- 8. How are connectivity and interoperability between client/server achieved?
- 9. One disadvantage of the Client/Server system is lack of control in a Database Management environment. Justify.
- 10. Explain the DBMS concept in client/server architecture.

## Part B

# (Answer any one question from each module. Each question carries 14 Marks)

- 11. (a) Differentiate between Transaction server and Data server system with examples. (7)
  - (b) Computing in client server architecture over Mainframe architecture has certain advantages and disadvantages. Describe at least three advantages (7) and disadvantages for each architecture.

#### OR

# 12. (a) Explain various Clients/Server system development tools. (6) (b) Classify and describe the driving forces that drive the move to Client/Server (8) computing. 13. (a) Explain the role of mainframe-centric model in Client/Server computing? (5) (b) Describe the three types of Client/Server systems in existence (9) OR 14. (a) List and explain the general forces behind the architecture for business (7) information systems (b) Explain the different distribution styles. (7) 15. (a) Illustrate the concept of rightsizing and downsizing in Client/Server (7) Computing

(b) What is client server system development methodology? Explain the (7)

different phases of System Integration Life-Cycle.

# OR

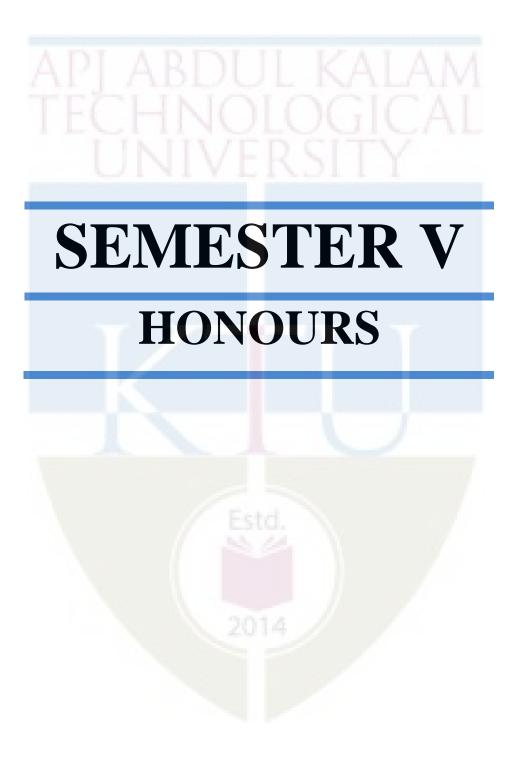
16.	(a)	In Client/Server computing, explain the following with examples i. Dynamic Data Exchange ii. RPC, Remote Procedure Call iii. Remote Boot Service iv. Diskless Computer v. Object-linking and embedding	(10)
	(b)	Explain the functions and features of Network Operating System	(4)
17.	(a)	Explain about network management and remote system management. How can security be provided to the network ?	(10)
	(b)	In client server architecture, what do you mean by Availability, Reliability, Serviceability and Security? Explain with examples.	(4)
		OR	
18.	(a)	Client server is modular infrastructure, this is intended to improve Usability, Flexibility, Interoperability and Scalability. Explain each term with an example, in each case how it helps to improve the functionality of client server architecture.	(7)
	(b)	Explain about network management and remote system management. How can security be provided to network?	(7)
19.	(a)	Explain the different types of Client/Server Database Architecture	(9)
	(b)	List and explain the main components of Database middleware	(5)
		OR	
20.	(a)	Discuss types of database utilities, tools and their functions	(7)
	(b)	Discuss about the role of traditional and web databases in handling client/server based applications.	(7)

# **Teaching Plan**

	Module- 1( Introduction)					
1.1	Basic Client/Server Computing Model	1 hour				
1.2	Server for Every Client- File Server, Print Server	1 hour				
1.3	Application Server, Mail Server, Directory Services Server					
1.4	Web Server, Database Server	1 hour				
1.5	Transaction Servers	1 hour				
1.6	Client/Server-Fat or Thin	1 hour				
1.7	Stateless or Stateful	1 hour				
1.8	Servers and Mainframes	1 hour				
1.9	Client/Server Functions	1 hour				
1.1 0	Driving Forces behind Client/Server Computing- Business Perspective, Technology Perspective					
	Module- 2 (Client/Server Classification)	(10 hours)				
2.1	Client/Server Types-Single Client/Single Server	1 hour				
2.2	Multiple Clients/Single Server, Multiple Clients/Multiple Servers	1 hour				
2.3	Integration With Distributed Computing	1 hour				
2.4	Alternatives To Client/Server Systems	1 hour				
2.5	Classification of Client/Server Systems- Two-Tier Computing, Middleware	1 hour				
2.6	Three-Tier Computing- Model View Controller (MVC)	1 hour				
2.7	Principles behind Client/Server Systems.	1 hour				
2.8	Client/Server Topologies	1 hour				
2.9	Existing Client/Server Architecture					
2.10	Architecture for Business Information System	1 hour				
	Module -3 (Client/Server Application Components)	(9 hours)				
3.1	The client: Services, Request for services, RPC	1 hour				
3.2	Windows services, Print services, Remote boot services	1 hour				

# B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

3.3	Utility Services & Other Services	1 hour			
3.4	Server- Detailed server functionality, Network operating system	1 hour			
3.5	Available platforms, Server operating system	1 hour			
3.6	Organizational Expectations, Improving performance of client/server applications				
3.7	Single system image, Downsizing and Rightsizing	1 hour			
3.8	Advantages and disadvantages of Client/Server computing	1 hour			
3.9	Applications of Client/Server	1 hour			
	Module -4 (Client/ Server Systems Services and Support)	(8 hours)			
4.1	Services and Support, System administration	1 hour			
4.2	Availability, Reliability	1 hour			
4.3	Scalability, Observability, Agility	1 hour			
4.4	Serviceability, Software Distribution	1 hour			
4.5	Performance	1 hour			
4.6	Network management	1 hour			
4.7	Remote Systems Management- RDP, Telnet, SSH	1 hour			
4.8	Security, LAN and Network Management issues	1 hour			
Mod	ule -5(Client/Server Technology and Databases)	(8 hours)			
5.1	Client/Server Technology and Databases - Storing Data	1 hour			
5.2	Database System Architectures	1 hour			
5.3	Client/Server In Respect Of Databases- Client/Server Databases	1 hour			
5.4	Client/Server Database Computing	1 hour			
5.5	Database Computing Vs. Mainframe, PC/File Server Computing	1 hour			
5.	Client/Server Database Architecture - Process-Per-Client Architecture	1 hour			
5.7	Multi-Threaded Architecture, Hybrid Architecture	1 hour			
5.8	Database Middleware Component - Application Programming Interface, Database Translator, Network Translator	1 hour			



CST 393	CRYPTOGRAPHIC ALGORITHMS	Category	L	Т	Р	Credit	Year of Introduction
393		VAC	3	1	0	4	2019
Preamble:							

# Preamble:

The course on Cryptographic Algorithms aims at exploring various algorithms deployed in offering confidentiality, integrity, authentication and non-repudiation services. This course covers classical encryption techniques, symmetric and public key crypto-system, key exchange and management, and authentication functions. The concepts covered in this course enable the learners in effective use of cryptographic algorithms for real life applications.

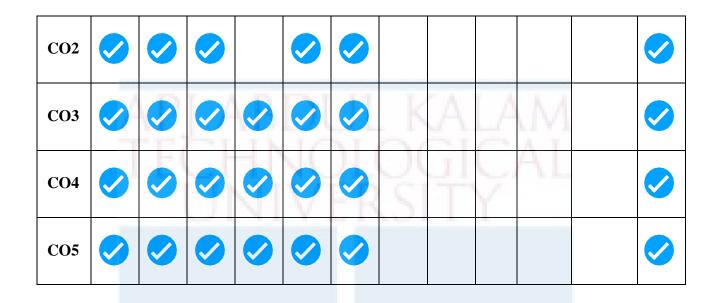
Prerequisite: A sound background in Number Theory.

Course Outcomes: After the completion of the course the student will be able to

CO1	Identify the security services provided for different types of security attacks. (Cognitive Knowledge Level : Understand)
CO2	Summarize the classical encryption techniques for information hiding. (Cognitive Knowledge Level: Apply)
CO3	Illustrate symmetric / asymmetric key cryptographic algorithms for secure communication.(Cognitive Knowledge Level: Apply)
CO4	Interpret key management techniques for secure communication.(Cognitive Knowledge Level: Understand)
CO5	Summarize message authentication functions in a secure communication scenario.(Cognitive Knowledge Level: Understand)

# Mapping of course outcomes with program outcomes

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO 9	PO1 0	PO11	PO1 2
CO1												



	Abstract POs defined by National Board of Accreditation								
PO#	Broad PO	PO#	Broad PO						
PO1	Engineering Knowledge	PO7	Environment and Sustainability						
PO2	Problem Analysis	PO8	Ethics						
PO3	Design/Development of solutions	PO9	Individual and team work						
PO4	Conduct investigations of complex problems	PO10	Communication						
PO5	Modern tool usage	PO11	Project Management and						
PO6	The Engineer and Society 2014	PO12	Life long learning						

# **Assessment Pattern**

Bloom's	Continuous Assessmer	nt Tests	End Semester
Category	Test1 (Percentage)	Test2 (Percent	Examinati on Marks

		age)	
Remember	30	30	30
Understand	30	30	30
Apply	40	40	40
Analyze	JNIVER	SITY	
Evaluate			
Create			

# **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3 hours

# **Continuous Internal Evaluation Pattern:**

Attendance : 10 marks

Continuous Assessment Tests : 25 marks

Continuous Assessment Assignment : 15 marks

# **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks.

First Internal Examination shall be preferably conducted after completing the first half of the syllabus and the Second Internal Examination shall be preferably conducted after completing remaining part of the syllabus.

There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly covered module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly covered module), each with 7 marks. Out of the 7 questions in Part B, a student should answer any 5.

# **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.

# **Syllabus**

# Module-1 (Introduction to the Concepts of Security)

Need for security, Security approaches, Principles of security, Types of attacks, OSI Security Architecture, Classical encryption techniques - Substitution techniques, Transposition techniques. Stream cipher, Block cipher, Public key cryptosystems vs. Symmetric key cryptosystems, Encrypting communication channels.

# Module-2 (Symmetric Key Cryptosystems)

Overview of symmetric key cryptography, Block cipher principles, Data Encryption Standard (DES), Differential and Linear cryptanalysis, Double DES, Triple DES, International Data Encryption Algorithm (IDEA), Advanced Encryption Algorithm (AES),Block cipher modes of operation, Stream cipher, RC4.

### Module-3 (Public Key Cryptosystems)

Principles of public key cryptosystems, RSA algorithm, RSA illustration, Attacks, ElGamal cryptographic system, Knapsack algorithm, Diffie-Hellman key exchange algorithm, Elliptical curve cryptosystems.

# Module-4 (Key Management)

Symmetric key distribution using symmetric encryption, Symmetric key distribution using asymmetric encryption, Distribution of public keys, Generating keys, transferring keys, Verifying keys, Updating keys, Storing keys, Backup keys, Compromised keys, Public key infrastructure.

# Module – 5 (Authentication)

Authentication requirements, Authentication functions, Message authentication codes (MAC), Hash functions, Security of Hash functions and MAC, Message Digest 5 (MD5), Secure Hash Algorithm (SHA)-512, Hash-based Message Authentication Code (HMAC), Cipher-based Message Authentication Code (CMAC), X.509 Authentication services.

# **Text Books**

- 1. William Stallings, Cryptography and Network Security Principles and Practice, Pearson Edu, 6e.
- 2. Bruice Schneier, Applied Cryptography Protocols, Algorithms and source code in C, Wiley,2e.

# References

- 1. Behrouz A. Forouzan, Cryptography and Network Security, McGraw Hill, 2e.
- 2. Johannes A. Buchmann, Introduction to Cryptography, Springer, 2e.

3. DouglasR. Stinson, Cryptography Theory and Practice, 3e,Chapman & Hall/CRC, 2006.

4. Bernard Menezes, Network Security and Cryptography, Cengage Learning, 2011.

# Sample Course Level Assessment Questions

# **Course Outcome 1 (CO1):**

- 1. Consider an automated teller machine (ATM) in which users provide a personal identification number (PIN) and a card for account access. Give examples of confidentiality, integrity, and availability requirements associated with the system and, in each case, indicate the degree of importance of the requirement.
- 2. Discuss the different security services provided for preventing security attacks.

# Course Outcome 2 (CO2):

1. The encryption key in a transposition cipher is (3,2,6,1,5,4). Find the decryption key

2. Discuss the process of encryption in Vernam cipher

# Course Outcome 3 (CO3):

1. Devise a meet-in-the-middle attack for a triple DES.

- 2. Write an algorithm for the InvSubBytes transformation and implement using python (Assignment)
- 3. Consider the following elliptic curve signature scheme. We have a global elliptic curve, prime p, and "generator" G. Alice picks a private signing key  $X_A$  and forms the public verifying  $Y_A = X_A G$ . To sign a message M:
  - Alice picks a value k
    - Alice sends Bob M, k and the signature  $S = M kX_AG$ .
    - Bob verifies that  $M=S+kY_A$ .

Show that the verification process produces an equality if the signature is valid.

- **4.** Write an algorithm to add two points on an elliptic curve over GF(*p*) and implement using Python. (Assignment)
- 5. Write an algorithm for encryption using knapsack cryptosystem and implement using Java. (Assignment)

# **Course Outcome4 (CO4):**

- 1. List four general categories of schemes for the distribution of public keys.
- 2. What are the essential ingredients of a public-key directory?

# **Course Outcome 5 (CO5):**

- 1. State the value of the length field in SHA-512 if the length of the message is 1919 bits and 1920 bits.
- 2. Write an algorithm in pseudo code for HMAC and implement using Python (Assignment)

### **Model Question Paper**

# **QP CODE:**

Reg No:

Name :

PAGES:3

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

# FIFTH SEMESTER B.TECH DEGREE EXAMINATION(HONORS), MONTH & YEAR

### **Course Code: CST 393**

### **Course Name: Cryptographic Algorithms**

Max.Marks:100

**Duration: 3 Hours** 

### PART A

# Answer all Questions. Each question carries 3 Marks

- 1. State the two approaches in attacking a cipher.
- 2. Define Substitution Cipher. Encrypt using one time pad M = HONORS and K = CIPHER.
- 3. Specify the purpose of S-Boxes in Data Encryption Standard (DES).
- 4. Differentiate between diffusion and confusion.
- 5. Perform encryption using RSA Algorithm for the following p=7; q=11; e=13; M=5.
- 6. Is Diffie-Hellman key exchange protocol vulnerable? Justify.
- 7. List the techniques for distribution of public keys.
- 8. Define a certificate authority and its relation to public key cryptography.
- 9. Distinguish between integrity and message authentication.
- 10. What types of attacks are addressed by message authentication?

(6)

10

(4)

### Part B

# (Answer any one question from each module. Each question carries 14 Marks)

11 (a)		
11. (a)	With a work shot she Erstein OSI Constitute as hits from and all	
	With a neat sketch, Explain OSI Security architecture model.	
	I IN TH /FD CITY/	(8)

(b) How does link encryption differ from end-to-end encryption? Explain.

# OR

- 12. (a) Encrypt the text "cryptography" using the Hill Cipher with the key  $\begin{pmatrix} 9 & 4 \\ 5 & 7 \end{pmatrix}$ . Show the calculations. (8)
  - (b) Illustrate the steps involved in encrypting a plain text using playfair cipher with an example.
- 13. (a) With a neat sketch, explain a single round in DES.
  - (b) Explain encryption and decryption using 2 keys and 3 keys of triple DES.

### OR

# 14. (a) Explain the block cipher modes i) Cipher feedback mode ii) Output (8) feedback mode. (b) Describe the four types of transformations in AES. (6) 15. (a) Write an algorithm for generating public and private key using Elliptical (10) curve cryptography.

	(b)	The equation $y^2=x^3 +x+1$ , the calculation is done modulo 13. Add two points R= P+Q, where P= (4,2) and Q= (10,6).	(4)
		APJ ABDUP KALAM	
16.		User A and B use the Diffie-Hellman key exchange technique with a common prime $q=71$ and primitive root alpha=7.	
	(a)	If user A has private key $X_A = 3$ , What is A's public key $Y_A$ ?	(7)
	(b)	If user B has private key $X_B = 6$ , What is A's public key $Y_B$ ?	(7)
17.	(a)	Define a session key and show how a KDC can create can create a session key between Alice and Bob.	(7)
	(b)	What are the requirements for the use of a public-key certificate scheme?	(7)
		OR	
18.	(a)	What are the core components of a PKI? Briefly describe each component.	(8)
	(b)	Describe the following (i) Updating keys (ii) Compromised Keys.	(6)
19.	(a)	Describe how SHA-512 logic produce message digest	(10)
	(b)	Distinguish between HMAC and CMAC	(4)
		201 OR	
20.	(a)	Specify the format for X.509 certificate. Explain the steps required to obtain user's certificate.	(7)
	(b)	With suitable block diagrams, explain the types of functions that may beused to produce an authenticator.	(8)

# **Teaching Plan**

No	Contents	No of Lecture Hrs						
	Module - 1 (Introduction to the Concepts of Security) (9 hrs)							
1.1	Need for security, Security approaches	1 hour						
1.2	Principles of security, Types of attacks	1 hour						
1.3	OSI Security Architecture	1 hour						
1.4	Classical encryption techniques: Substitution techniques(Caesar cipher, Monoalphabetic cipher, Playfair cipher)	1 hour						
1.5	Classical encryption techniques: Substitution techniques (Hill cipher, Polyalphabetic cipher, One-time pad)	1 hour						
1.6	Classical encryption techniques: Transposition techniques	1 hour						
1.7	Stream cipher, Block cipher	1 hour						
1.8	Public- key cryptosystems vs. Symmetric key cryptosystems	1 hour						
1.9	Encrypting communication channels	1 hour						
	Module - 2 (Symmetric key cryptosystems) (11 hrs)							
2.1	Overview of symmetric key cryptography	1 hour						
2.2	Block cipher principles	1 hour						
2.3	Data Encryption Standard (DES)	1 hour						
2.4	DES design criteria	1 hour						
2.5	Differential and Linear cryptanalysis	1 hour						
2.6	Double DES, Triple DES	1 hour						

2.7	IDEA	1 hour				
2.8	.8 Advanced Encryption Algorithm (AES structure)					
2.9	Advanced Encryption Algorithm (Transformations)	1 hour				
2.10	Block cipher modes of operation	1 hour				
2.11	Stream cipher, RC4	1 hour				
	Module - 3 (Public key cryptosystems) (8 hrs)					
3.1	Principles of public key cryptosystems	1 hour				
3.2	RSA algorithm	1 hour				
3.3	RSA illustration, Attacks	1 hour				
3.4	ElGamal cryptographic system	1 hour				
3.5	Knapsack algorithm	1 hour				
3.6	Diffie-Hellman key exchange algorithm	1 hour				
3.7	Elliptical curve cryptosystems(Elliptical curve arithmetic)	1 hour				
3.8	Elliptical curve cryptosystems (Elliptical curve algorithm)	1 hour				
	Module - 4 (Key Management) (8 hrs) [Text book-2]					
4.1	Symmetric key distribution using symmetric encryption	1 hour				
4.2	Symmetric key distribution using asymmetric encryption	1 hour				
4.3	Distribution of public keys	1 hour				
4.4	Generating keys, Transferring keys	1 hour				

4.5	Verifying keys, Updating keys	1 hour
4.6	Storing keys, Backup keys	1 hour
4.7	Compromised keys	1 hour
4.8	Public key infrastructure	1 hour
	Module - 5 (Authentication) (9 hrs)	
5.1	Authentication requirements	1 hour
5.2	Authentication functions	1 hour
5.3	Message Authentication Codes (MAC)	1 hour
5.4	Hash functions	1 hour
5.5	Security of Hash functions and MAC	1 hour
5.6	MD5	1 hour
5.7	SHA-512	1 hour
5.8	HMAC, CMAC	1 hour
5.9	X.509 Authentication services	1 hour

		CATEGORY	L	Τ	Р	Credit
AIT 395	COMPUTATIONAL BIOLOGY	Honors	3	1	0	4

# **Preamble:**

This course helps the learners to understand concepts in Genomics, Proteomics Computational Biology, Next Generation Sequencing, NGS Data Analysis and Systems biology. It enables the learners to understand various Next Generation Sequencing Techniques, analysis and interpretation of the NGS Data. Also, course introduces computational and mathematical analysis and modeling of complex biological systems and Systems Biology

**Prerequisite:** Basic background in Bioinformatics

Course Outcomes: After the completion of the course, the student will be able to

CO 1	Describe the basic concepts of genomics, microarray, protein structure determination
	and prediction(Cognitive knowledge level: Understand)
CO 2	Explain the fundamental aspects drug discovery and molecular modelling (Cognitive knowledge level: Apply)
CO 3	Demonstrate Networks in Biology, types of networks and its representation (Cognitive knowledge level : Apply)
~~ .	
CO 4	Explain Next Generation sequencing Technologies and DNA Protein interaction analysis(Cognitive knowledge level: Understand)

# Mapping of course outcomes with program outcomes

PO1	PO2	PO3	PO4	PO5	PO6	<b>PO7</b>	PO8	PO9	PO10	PO11	PO12
0	$\checkmark$				201						$\bigcirc$
0	0	0	0	0			-				<ul> <li>Image: Construction</li> </ul>
0	0	0	<ul> <li>Image: Construction</li> </ul>	0							<ul> <li>Image: Construction</li> </ul>
$\oslash$	$\oslash$	$\oslash$	Ø	$\oslash$							Ø
0	0			<ul> <li>Image: Construction</li> </ul>							<ul> <li>Image: Construction</li> </ul>
	0 0 0	<ul> <li>Ø</li> <li>√</li> <li>Ø</li> <li>Ø</li> <li>Ø</li> <li>Ø</li> <li>Ø</li> <li>Ø</li> <li>Ø</li> </ul>	$ \begin{array}{c c}                                    $	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$		$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$

PO#	Broad PO	PO#	Broad PO
PO1	Engineering Knowledge	PO7	Environment and Sustainability
PO2	Problem Analysis	PO8	Ethics
PO3	Design/Development of solutions	PO9	Individual and team work
PO4	Conduct investigations of complex problems	PO10	Communication
PO5	Modern tool usage	PO11	Project Management and Finance
PO6	The Engineer and Society	PO12	Life long learning

# Assessment Pattern

Bloom's Category		Continuous Asse	essment Tests	End Semester Examination
		<b>Test1 (%)</b>	Test2 (%)	_
Remember		30	30	30
Understand		50	50	50
Apply		20	20	20
Analyse				
Evaluate				
Create				

# Mark Distribution

Total	CIE	ESE Marks	ESE
Marks	Marks		Duration
150	50	100	3

# **Continuous Internal Evaluation Pattern:**

Attendance	10 marks
Continuous Assessment Tests (Average of Series Tests 1& 2)	25 marks
Continuous Assessment Assignment	15 marks

### **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed module), each with 7 marks. Out of the 7 questions, a student should answer any5.

### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 full questions from each module of which student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.



# SYLLABUS`

### Module -01 (Genomics and Proteomics)

Genes, Genes in genomes, Genomes of prokaryotes and Eukaryotes, Protein-coding genes, RNA, Single-nucleotide polymorphisms, Microarray, Analysis of microarray data, Proteins and peptides, Experimental Protein structure identification, computational methods for protein structure prediction, Homology modelling, Protein folding and fold recognition.

# Module-02 (Computer Aided Drug Discovery)

Drug discovery pipeline, Drug target identification & validation, Active site identification, pharmacophore, Lead/Ligand identification, lead compound optimization, Binding energy calculation, Energy Minimization. Molecular modelling in drug discovery, concept of Molecular Dynamics, concept of Absorption, Distribution, Metabolism and Excretion (ADME), Quantitative Structure-Activity Relationships.

# Module-03 (Network Biology)

Transcriptional Regulatory Networks, Genes and DNA Regulatory Regions, Genetic Interaction Map, Protein Interaction Networks, Experimental methodologies to obtain Protein Interaction Data, Computational methods to Predict Protein-Protein Interactions, Visualization of Protein Interaction Networks, Metabolic Networks, Interacting Partners, Mathematical Representation

# Module-04 (Next Generation Sequencing and analysis)

A Typical NGS Experimental Workflow, Next-Generation Sequencing (NGS) Technologies, Illumina Reversible Dye-Terminator Sequencing, Ion Torrent Semiconductor Sequencing, Pacific Biosciences Single Molecule Real-Time (SMRT) Sequencing, RNA-sequencing (RNA Seq), Protein-DNA Interaction Analysis (ChIP-Seq)

# Module-05 (NGS Data Analysis)

Base Calling, FASTQ File Format, and Base Quality Score, NGS Data Quality Control and Preprocessing, Reads Mapping, Mapping Approaches and Algorithms, Selection of Mapping Algorithms and Reference Genome Sequences, SAM/BAM as the Standard Mapping File Format, Mapping File Examination and Operation, Tertiary Analysis

# Books

- 1. Lesk, Arthur M. Introduction to Bioinformatics. United Kingdom, Oxford University Press, 2019.
- 2. Biological Networks. Singapore, World Scientific Publishing Company, 2007.
- 3. Wang, Xinkun. Next-Generation Sequencing Data Analysis. United States, CRC Press, 2016.

# References

1. Tiwary, Basant K. Bioinformatics and Computational Biology: A Primer for Biologists. Singapore, Springer Singapore, 2021.

- 2. Benfey, Philip N. Quickstart Molecular Biology: An Introductory Course for Mathematicians, Physicists, and Computational Scientists. United States, Cold Spring Harbor Laboratory Press, 2014.
- 3. Baxevanis, Andreas D., Gary D. Bader, and David S. Wishart, eds. *Bioinformatics*. John Wiley & Sons, 2020.
- 4. Shaik, Noor Ahmad, et al. Essentials of Bioinformatics, Volume I. Springer, 2019
- 5. Selzer, Paul M., Richard J. Marhöfer, and Andreas Rohwer, *Applied bioinformatics*. *An introduction–Springer, Verlag*, 2008.
- 6. S C Rastogi, N Mendiratta and P Rastogi, *Bioinformatics: Methods and Applications*, PHI Learning Private Limited, New Delhi, 2015.
- 7. D E Krane and M L Raymer, *Fundamental Concepts of Bioinformatics*, Pearson Education, 2006.

# **Course Level Assessment Questions**

# **Course Outcome 1 (CO1):**

- 1. Compare and contrast the genomes of Prokaryotes and Eukaryotes
- 2. Summarize the method of DNA microarray and its analysis.
- 3. Using the online tool SWISS-MODEL, develop model of Homo sapiens (Human) Leptin protein and interpret your result

# **Course Outcome 2 (CO2):**

- 1. Explain the process of computer aided drug discovery and various step involved in it
- 2. Explain the process of molecular modelling in drug discovery

# **Course Outcome 3 (CO3):**

- 1. Differentiate between Transcriptional and protein interaction networks
- 2. From the STRING database identify the interactions of Homo sapiens TP53 protein and interpret your result

# **Course Outcome 4 (CO4):**

- 1. Summarize Next Generation Sequencing methods.
- 2. Explain The Protein- DNA interaction analysis with the help of ChIP-Seq
- 3. What can RNA-seq reveal?

# **Course Outcome 5 (CO5):**

- 1. Illustrate the process involved in Data Quality control and preprocessing in Next Generation Sequencing
- 2. Explain the mapping algorithms and reference genome sequences

# **TEACHING PLAN**

No	Contents	No of Lecture (45Hrs)
	Module -01 (Genomics and Phylogenetics) (9hrs)	i
1.1	Genes, Genes in genomes.	1
1.2	Genomes of prokaryotes and Eukaryotes	1
1.3	Protein-coding genes, RNA, Single-nucleotide polymorphisms	A 1
1.4	Microarrays	1
1.5	Analysis of microarray data	1
1.6	Proteins and peptides	1
1.7	Experimental Protein structure identification	1
1.8	Computational methods for protein structure prediction	1
1.9	Homology modelling, Protein folding and fold recognition	1
	Module-02 (Computer Aided Drug Discovery)(9hrs)	
2.1	Drug discovery pipeline	1
2.2	Drug target identification & validation	1
2.3	Active site identification, pharmacophore	1
2.4	Lead/Ligand identification	1
2.5	lead compound optimization, Binding energy calculation, Energy Minimization	1
2.6	Molecular modelling in drug discovery	1
2.7	Concept of Molecular Dynamics	1
2.8	Concept of Absorption, Distribution, Metabolism and Excretion (ADME)	1
2.9	Quantitative Structure-Activity Relationship	1
	Module-03 (Network Biology)(9hrs)	
3.1	Transcriptional Regulatory Networks	1
3.2	Genes and DNA Regulatory Regions,	1
3.3	Genetic Interaction Map,	1
3.4	Protein Interaction Networks	1
3.5	Experimental methodologies to obtain Protein Interaction Data	1
3.6	Computational methods to Predict Protein-Protein Interactions	1
3.7	Visualization of Protein Interaction Networks	1
3.8	Metabolic Networks- Interacting Partners	
3.9	Metabolic Networks- Mathematical Representation	

4.1	A Typical NGS Experimental Workflow	1
4.2	Next-Generation Sequencing (NGS) Technologies	1
4.3	Next-Generation Sequencing (NGS) Technologies	1
4.4	Illumina Reversible Dye-Terminator Sequencing	1
4.5	Ion Torrent Semiconductor Sequencing	1
4.6	Pacific Biosciences Single Molecule Real-Time (SMRT) Sequencing	1
4.7	RNA-sequencing (RNA Seq)	1
4.8	Protein-DNA Interaction Analysis (ChIP-Seq)	1
5.1	NGS data,FASTQ File Format	1
	Module-05 (NGS Data Analysis)(10hrs)	
5.2	Base Calling, Base Quality Score	1
5.3	NGS Data Quality Control	1
5.4	NGS data Preprocessing	1
5.5	Reads Mapping, Mapping Approaches and Algorithms,	1
5.6	Selection of Mapping Algorithms and Reference Genome Sequences	1
5.7	SAM/BAM as the Standard Mapping	1
5.7	Mapping File Examination and Operation	1
5.8 5.9	Tertiary Analysis	1

# **Model Question Paper**

**QP CODE:** 

Reg No: \_\_\_\_\_

Name: \_\_\_\_\_

PAGES: 4

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

# EIGHTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

Course Code: AIT 395

# **Course Name: COMPUTATIONAL BIOLOGY**

Max. Marks: 100

**Duration: 3 Hours** 

(10x3=30)

# PART A

# Answer All Questions. Each Question Carries 3 Marks

- 1. Distinguish between Genes, Genes in genomes.
- 2. What are the structural features of Eukaryotic cells?
- 3. What are SNPs and why are they important?
- 4. How do you identify the active site of a protein?
- 5. What is protein energy minimization?
- 6. List any three types of biochemical networks with one line description
- 7. What are reversible Dye-Terminators in NGS sequencing?
- 8. What is the difference between the DNA sent for Whole Exome sequencing vs ChIP sequencing?
- 9. List any three features of FastQ file format.
- 10. What is SAM format? How is BAM different from SAM?

# Part B

# (Answer any one question from each module. Each question carries 14 Marks)

- 11. (a) With the help of a neat diagram, explain a prokaryotic gene structure. Is a promoter at the upstream or downstream of a transcription unit? (7)
  - (b) What is homology modeling? Discuss the steps involved in the same (7)

# OR

12. (a)	Explain the design of a microarray experiment, detailing the various phases.	(7)
(b)	What experimental method is used to determine the tertiary protein structure? What are the computational methods?	(7)
13. (a)	Illustrate the computational drug discovery pipeline with a suitable flowchart	(7)

	(b)	What is Molecular modeling in drug discovery? Explain the process of molecular modelling.	(7)
		OR	
14.	(a)	Explain the scoring functions in molecular docking.	(7)
	(b)	Explain lead compound optimization, Binding energy calculation, Energy Minimization in the process of Computer aided drug discovery	(7)
15.	(a)	What is transcriptional control and why is it important? Explain how transcriptional regulatory networks plays an important role in gene expression and control?	(7)
	(b)	Explain how the computational methods helps in identifying the Protein– Protein Interactions	(7)
		OR	
16.	(a)	How the Protein–Protein Interactions are identified by using experimental methods.	(7)
	(b)	What is metabolic network? What are type of data are needed for metabolic network reconstruction?	(7)
17.	(a)	Explain any two next-generation sequencing techniques with their steps.	(7)
	(b)	How do you interpret a FastQC report?	(7)
		OR	
18.	(a)	What are the steps in RNA sequencing? Why is RNA-seq better than microarrays?	(7)
	(b)	illustrate the steps involved in mapping protein-DNA interactions using ChIP-sequencing	(7)
19.	(a)	How do you interpret per base sequence quality?What is the purpose of mapping reads to a reference genome?	(7)
	(b)	Explain any three mapping algorithms for the NGS.	(7)
		OR	
20.	(a)	Illustrate steps involved in the NGS data Preprocessing and Quality Control	(7)
	(b)	Discuss the significance of NGS in clinical diagnosis.	(7)

AIT 397	ADVANCED CONCEPTS IN	Category	L	Т	Р	Credit
	COMPUTER VISION	HONORS	3	1	0	

### Preamble:

This course enables the learners to understand the advanced concepts in computer vision. The course covers the basics of image processing, imaging geometry, image segmentation, feature extraction, object recognition and classification and common applications of computer vision. This course helps the students to design solutions for complex real-life problems.

**Prerequisite:** A sound knowledge of Mathematics and concepts of any programming language.

### Mapping of course outcomes with program outcomes

C01	Illustrate the concepts of image formation and image model. (Cognitive Knowledge Level: Understand)
CO2	Demonstrate various feature extraction and edge detection techniques. (Cognitive Knowledge Level: Apply
CO3	Apply edge-based and region-based image segmentation techniques. (Cognitive Knowledge Level: Apply)
CO4	Understand and implement image recognition and classification methods. (Cognitive Knowledge Level: Apply)
CO5	Explain the various applications of computer vision. (Cognitive Knowledge Level: Understand)

# Mapping of course outcomes with program outcomes

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO1 0	PO11	PO1 2
CO1	0		0	LD LD	0	N'		~i	2	A T		
CO2	$\bigcirc$	٢	٢	۲	۲	0	Se	1	7	1L		$\bigcirc$
CO3	٢		٢	0	0	0	3		1			
CO4	0					0						
CO5	0					٢						$\bigcirc$

	Abstract POs defined by National Board of Accreditation							
PO#	Broad PO	PO#	Broad PO					
PO1	Engineering Knowledge	PO7	Environment and Sustainability					
PO2	Problem Analysis	PO8	Ethics					
PO3	Design/Development of solutions	PO9	Individual and team work					
PO4	Conduct investigations of complex problems	PO10	Communication					
PO5	Modern tool usage	PO11	Project Management and Finance					
PO6	The Engineer and Society	PO12	Life long learning					

# **Assessment Pattern**

Bloom' s	Continuous	End Semester Examination		
Catego ry	Test 1 (%)	Test 2 (%)	Marks (%)	
Remember	30	ER <sup>30</sup> SIT	3 0	
Understand	30	30	3 0	
Apply	40	40	4 0	
Analyze	TZ			
Evaluate	K			
Create				

# **Mark Distribution**

Total	CIE	ESE	ESE
Marks	Marks	Marks	Duration
150	50	100	3

# **Continuous Internal Evaluation Pattern:**

Attendance

10 marks

Continuous Assessment Tests(Average of Internal Tests 1 & 2)

25 marks

Continuous Assessment Assignment

### 15 marks

### **Internal Examination Pattern**

Each of the two internal examinations has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing the remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed modules and 1 question from the should answer all questions, a student should answer any 5.

### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 full questions from each module of which student should answer any one. Each question can have a maximum of 2 subdivisions and carries 14 marks.

# Syllabus

# Module – 1 (Image Formation and Processing)

Image formation and Image model- Components of a vision system- Cameras- camera model and camera calibration-Radiometry- Light in space- Light in surface - Sources, shadows and shading.

Fundamentals of Image processing: Basic steps of Image processing system sampling and quantization of an Image – Basic relationship between pixels.

# **Module - 2(Feature Extraction)**

**Points and Patches** – Feature detectors, feature descriptors, feature matching, feature tracking. **Edges** – edge detection, edge linking. **Lines** - Successive approximation, Hough transforms, Vanishing points.

### Module - 3 (Image Segmentation)

Classification of segmentation techniques, Edge detection, Edge linking, Thresholding, Region

growing, Region splitting and merging, Watershed based segmentation. Shadow detection and removal. Image processing using OpenCV - blending, smoothing, and reshaping.

# Module - 4 (Image Recognition and Classification)

Shape based object classification, Motion based object classification, Viola Jones Object Detection Framework, Object classification using CNNs, use of RCNN for object classification.

# **Module** - 5 (Applications)

Speech and Handwriting Recognition, Automatic Face Recognition, Video Segmentation and Keyframe Extraction, Real-Time Hand Pose Recognition.

# **Text Books**

- 1. David A. Forsyth & Jean Ponce, Computer vision A Modern Approach, Prentice Hall, 2002.
- 2. Richard Szeliski, Computer Vision: Algorithms and Applications, Springer.
- 3. Maheshkumar H Kolekar, "Intelligent Video Surveillance Systems: An Algorithmic Approach", CRC Press.
- 4. Francesco Camastra, Alessandro Vinciarelli, "Machine Learning for Audio, Image and Video Analysis: Theory and Applications", Springer 2015.

# **Reference Books**

- 1. Reinhard Klette, "Concise Computer Vision: An Introduction into Theory and Algorithms", Springer London, 2014.
- 2. Olivier Faugeras, "Three-Dimensional Computer Vision", The MIT Press, 1993.

# **Course Level Assessment Questions**

### **Course Outcome1 (CO1):**

- 1. Explain the components of a visual system.
- 2. Elaborate on the image formation model.

# **Course Outcome 2(CO2):**

1. Explain edge linking through Hough Transform.

2. Discuss how feature extraction is done in image processing.

### Course Outcome 3(CO3):

- 1. Compare the following methods for image segmentation: a) multiple thresholding, b) global thresholding c) local thresholding.
- 2. Justify the role of region growing, region splitting and region merging operations in any of the computer vision applications.

### **Course Outcome 4(CO4):** .

- 1. Explain convolution stage and pooling stage of a typical CNN layer.
- 2. Illustrate Viola Jones object detection framework.

### **Course Outcome 5(CO5):**

- 1. Elaborate on how computer vision helps in automatic face recognition applications.
- 2. Discuss how computer vision helps in tackling complex real world problems.



# **Model Question Paper**

<b>QP CODE:</b>		
Reg No:	API ABDUL KALAM	
Name:	TECHNOLOGICAL	PAGES : 3
	APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY	

### FIFTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

Course Code: AIT 397

**Course Name: Advanced Concepts in Computer Vision** 

Max. Marks: 100

**Duration: 3 Hours** 

# PART A

# Answer All Questions. Each Question Carries 3 Marks

- 1. Explain the working of a pinhole camera, Derive the expression for pinhole perspective projection.
- 2. Illustrate "foreshortening" with a neat diagram.
- 3. Explain edge linking through Hough Transform.
- 4. Illustrate any two techniques for vanishing point detection in an image.
- 5. Compare following methods for image segmentation a, multiple thresholding, b, global thresholding c, local thresholding.
- 6. Draw the flowchart of foreground-pixel extraction by edge-based shadow removal
- 7. Why is a convolutional neural network preferred over a dense neural network for an image classification task?
- 8. Assess the relevance of selective search algorithm in RCNN for object classification

- 9. Draw the diagram which shows the general scheme of a recognition system.
- 10. Illustrate steps in feature extraction from handwritten images.

# Part B

(10x3=30)

# (Answer any one question from each module. Each question carries 14 Marks)

- 11. (a) State different limitations of pinhole cameras and how to overcome these limitations. (9)
  - (b) What are shadows? Differentiate umbra from penumbra. How is a self shadow(5) different from a cast shadow?

### OR

12. (a)	Explain the local shading model. How are area sources different from line sources?	(7)
(b)	Define Camera Calibration. Explain intrinsic and extrinsic parameters of a camera.	(7)
<b>13.</b> (a)	Assess the role of adaptive non-maximal suppression (ANMS) in feature detection.	(4)
(b)	Illustrate following techniques:	(10)

- i) Bias and gain normalization (MOPS).
- ii) Gradient location-orientation histogram (GLOH)

### OR

<b>14.</b> (a)	Illustrate any 2 techniques in Successive approximation.	(4)
(b)	Compare Scale invariant feature transform (SIFT) and PCA-SIFT.	(5)
<b>15.</b> (a)	Illustrate Gradient operator and Laplacian operator with one example for each.	(10)
(b)	Illustrate Watershed Algorithms.	(4)

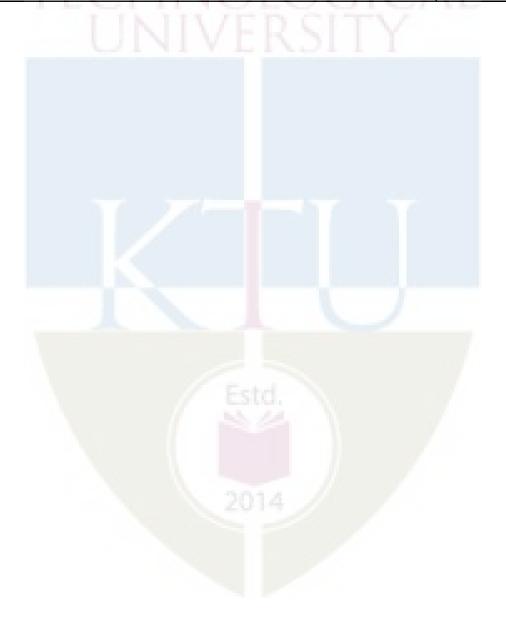
16.	(a)	With the help of a diagram illustrate region splitting and merging.	(7)
	(b)	Compare blending, smoothing, and reshaping functions using OpenCV.	(7)
17.	(a)	Differentiate between convolution stage and pooling stage of a typical CNN layer.	(8)
	(b)	Assess the role of dispersedness in shape based object classification.	(6)
		OR	
18.	(a)	Illustrate Viola Jones object detection framework.	(8)
	(b)	Explain the steps in motion based object classification.	(6)
19.	(a)	Illustrate shot boundary detection through pixel-based approaches and block-based approaches .	(7)
	(b)	Explain different approaches in keyframe extraction problems.	(7)
		OR	
20.	(a)	Illustrate shot boundary detection through histogram-based approaches and clustering-based approaches.	(6)
	(b)	Illustrate HMM training in speech and handwriting recognition.	(8)

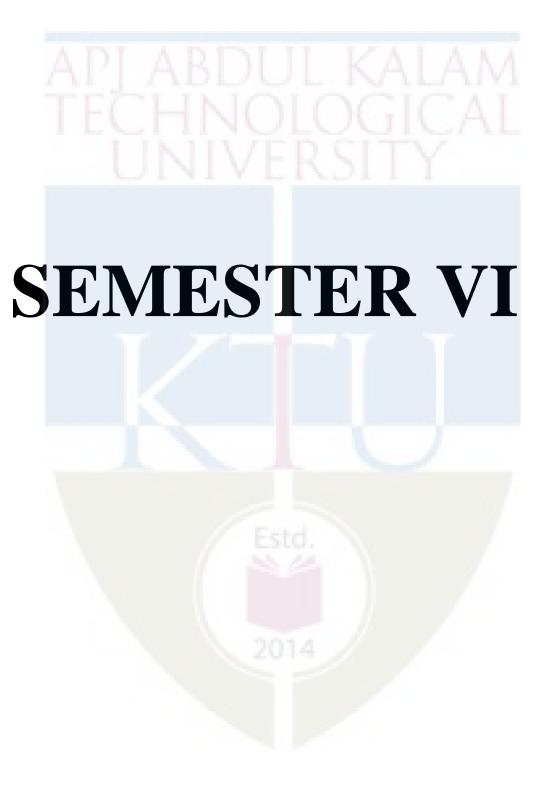
# **Teaching Plan**

No	Contents	No. of Lecture Hours (42 hrs)	
	Module – 1 (Image Formation and Processing) (8 hours)	M	
1.1	Image formation and Image model-Introduction	1 hour	
1.2	Components of a vision system- Cameras-Camera model	1 hour	
1.3	Camera calibration	1 hour	
1.4	Radiometry- Light in space-Light in surface	1 hour	
1.5	Sources-Shadows and shading	1 hour	
1.6	Fundamentals of Image processing: Basic steps of Image processing system	1 hour	
1.7	Sampling and quantization of an Image	1 hour	
1.8	Basic relationship between pixels.	1 hour	
	Module-2( Feature Extraction) (8 hours)		
2.1	Points and Patches – Feature detectors	1 hour	
2.2	Feature descriptors 2014	1 hour	
2.3	Feature matching	1 hour	
2.4	Feature tracking.	1 hour	
2.5	Edges – edge detection, edge linking.	1 hour	
2.6	Lines - Successive approximation	1 hour	

2.7	Hough transforms	1 hour
2.8	Vanishing points	1 hour
	Module-3( Image Segmentation)(9 hours)	M
3.1	Classification of segmentation techniques, Edge detection	1 hour
3.2	Edge linking	1 hour
3.3	Thresholding, Region growing	2 hours
3.4	Region splitting and merging	1 hour
3.5	Watershed based segmentation.	1 hour
3.6	Shadow detection and removal	1 hour
3.7	Image processing using OpenCV - blending	1 hour
3.8	Smoothing, and reshaping	1 hour
	Module-4(Image Recognition and Classification) (9 ho	ours)
4.1	Shape based object classification	1 hour
4.2	Motion based object classification	2 hours
4.3	Viola Jones Object Detection Framework	2 hours
4.4	Object classification using CNNs	2 hours
4.6	Use of RCNN for object classification.	2 hours
	Module-5( Applications)(8 hours)	
5.1	Speech and Handwriting Recognition	1 hour
5.2	Handwriting Recognition	1 hour

5.3	Automatic Face Recognition	2 hours
5.4	Video Segmentation	2 hours
5.5	Keyframe Extraction	1 hour
5.6	Real-Time Hand Pose Recognition.	1 hour





AMT 302	CONCEPTS IN NATURAL	Category	L	T	Р	Credit	Year of Introduction
	LANGUAGE PROCESSING	PEC	3	1	0	4	2019

# Preamble:

This course enables the learners to understand the concepts of Natural Language Processing. The course covers basic pre-processing steps, language models, text classification using machine learning algorithms, information and relation extraction methods, Information Retrieval, Question Answer Systems and Machine Translation models. This course enables the students to apply techniques and methods to solve challenging real-world problems in NLP.

# Prerequisite: Nil.

# Mapping of course outcomes with program outcomes

CO1	Summarize basic concepts and learning methods for NLP (Cognitive Knowledge Level: Understand)
CO2	Demonstrate the relevance of pre-processing methods on text data( <b>Cognitive</b> <b>Knowledge Level: Apply</b> )
CO3	Compare different language modelling techniques(Cognitive Knowledge Level: Apply)
CO4	

# Mapping of course outcomes with program outcomes

	PO1	PO2	PO3	PO4	PO5	PO6	P07	PO8	PO9	PO10	PO11	PO12
CO1								1				$\bigcirc$
CO2												$\bigcirc$
CO3												$\bigcirc$

# BTECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

CO4							$\bigotimes$
CO5	$\bigcirc$						

	Abstract POs defined by National Board of Accreditation							
PO#	Broad PO	PO#	Broad PO					
PO1	Engineering Knowledge	PO7	Environment and Sustainability					
PO2	Problem Analysis	PO8	Ethics					
PO3	Design/Development of solutions	PO9	Individual and team work					
PO4	Conduct investigations of complex problems	PO10	Communication					
PO5	Modern tool usage	PO11	Project Management and Finance					
PO6	The Engineer and Society	PO12	Life long learning					

# Assessment Pattern

Bloom's Category	Continuou	as Assessment Tests	End Semester Examination Marks (%)
Category	Test 1 (%)	Test 2 (%)	
Remember	30	E51C30	30
Understand	40	40	40
Apply	30	30	30
Analyze		2014	
Evaluate			
Create			

# **Mark Distribution**

10 marks

25 marks

15 marks

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3

<b>Continuous Internal Evaluation Patter</b>
--

Attendance

Continuous Assessment Tests (Average of InternalTests 1 & 2)

Continuous Assessment Assignment

# **Internal Examination Pattern**

Each of the two internal examinations has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed modules and 1 question from the partly should answer all questions, a student should answer any 5.

# **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 full questions from each module of which student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.

# Syllabus

# Module – 1 (Introduction to NLP)

NLP Tasks and Applications, Language-Building Blocks, Challenges of NLP, Machine Learning for NLP – Naïve Bayes Classifier, Logistic Regression, Support Vector Machines, Approaches to NLP-- Heuristics-Based NLP, Machine Learning-based NLP.

# Module - 2(Pre-processing and Representation Models)

NLP System Pipeline--Steps--Data Acquisition, Text Extraction and Clean-up, Pre-processing, Feature Engineering, Modelling, Evaluation, Post-Modelling Phases

Text Representation--Vector Space Models--Basic Vectorization Approaches--One-Hot Encoding, Bag of Words, Bag of N-Grams TF-IDF; Distributed Representations-- Word Embeddings, Doc2Vec.

# Module - 3 ( Classification and Information Extraction)

Text Classification--Text classification applications – Pipeline for building text classification systems, Naïve Bayes for Sentiment Classification – Naïve Bayes Classifier Training – Optimizing for Sentiment Analysis, Logistic Regression, Support Vector Machine for Text Classification

Information Extraction(IE)—IE Applications – The General Pipeline for IE - Named Entity Recognition(NER), Ambiguity in Named Entity Recognition – NER as Sequence Labeling – Evaluation of NER.

# **Module** - 4 (**Relation Detection and Information Retrieval**)

Relation Detection and Classification – Supervised Learning Approaches to Relation Analysis – Lightly Supervised Approaches to Relation Analysis – Evaluation of Relation Analysis systems

Information Retrieval – Term weighting and document scoring – Inverted Index – Evaluation of Information Retrieval Systems.

# Module - 5 (QA Systems and Machine Translation )

Question-Answering Systems – Factoid Question Answering – Question Processing – Passage Retrieval – Answer Processing – Evaluation of Factoid Answers

Machine Translation – Why Machine Translation is Hard – Classical Machine Translation – Direct Translation – Transfer – Statistical Machine Translation- The Phrase based Translation model – Alignment in MT – Training Alignment Models – Symmetrizing Alignments for Phrasebased MT – Decoding for Phrase-based Statistical MT

# **Text Books**

- 1. Daniel Jurafsky, James H. Martin, "Speech and Language Processing" (2<sup>nd</sup> and 3<sup>rd</sup> editions), Pearson Prentice Hall
- 2. SowmyaVajjala, BodhisattwaMajumder, Anuj Gupta, Harshit Surana," Practical Natural Language Processing: A Comprehensive Guide to Building Real-World NLP Systems" June 2020 Publisher(s): O'Reilly Media, Inc. ISBN: 9781492054054.

# ReferenceBooks

- 1. James Allen, "Natural Language Understanding", Second Edn, Pearson.
- 2. Christopher Manning and HinrichSchutze, Statistical Natural Language Processing, MIT Press.

# **Course Level Assessment Questions**

# **Course Outcome1 (CO1):**

- **1.** Explain the fundamental tasks that make up an NLP system.
- 2. Why is NLP considered a challenging problem domain?
- 3. The following table shows data about the profile of customers and whether they purchase computers or not. Given this data, use Naïve Bayes Classifier to classify the customer X (age = youth, income = medium, student = yes, credit rating = fair)

RID	age	income	student	credit_rating	Class: buys_computer
1	youth	high	no	fair	no
2	youth	high	no	excellent	no
3	middle_aged	high	no	fair	yes
4	senior	medium	no	fair	yes
5	senior	low	yes	fair	yes
6	senior	low	yes	excellent	no
7	middle_aged	low	yes	excellent	yes
8	youth	medium	no	fair	no
9	youth	low	yes	fair	yes
10	senior	medium	yes	fair	yes
11	youth	medium	yes	excel <mark>len</mark> t	yes
12	middle_aged	medium	no	excellent	yes
13	middle_aged	high	yes	fair	yes
14	senior	medium	no	excellent	no

**4.** Illustrate how linearly inseparable data can be made linearly separable by suitable mapping using kernel functions.

# **Course Outcome 2(CO2):**

- 1. Mention two issues associated with sentence segmentation.
- 2. Show how is lemmatization done using Python Library.
- Given a dataset of tweets, prepare the data for sentiment analysis by doing the following operations: conversion to lower casing, removal of punctuations, removal of stop-words, stemming, lemmatization, removal of emojis and removal of URLs. (Assignment Question)

# Course Outcome 3(CO3):

- 1. Compare Bag-of-Words model and Bag-of-n-gram model.
- 2. Illustrate how TF-IDF model is used to represent text. Mention the advantage of TF-IDF over other models.
- 3. A corpus of data is given below:
  - D1 Dog bites man.
  - D2 Man bites dog.
  - D3 Dog eats meat.
  - D4 Man eats food.

Use one hot-encoding and Bag-of-words models to represent "dog bites man".

4. Using the toy corpus given above, represent the sentence "Dog and Man eat meat" with TF-IDF model. Use python code for implementation. (Assignment Question)

## **Course Outcome 4(CO4):**

1. Given the following data about documents and contents, use tf-idf document scoring method to retrieve the document for the query "best game"

	The game was so exciting. The players excelled in every
Doc 1	department of the game.
Doc 2	It was an excellent game.
Doc 3	The game was not good. The moves were boring

- 2. A corpus of data is available from a social media platform that represents review of books. How can Naïve Bayes Classifier be used for sentiment analysis of the reviews? What changes can be made to this classifier to make it tuned for sentiment analysis.
- 3. Use python library to implement sentiment analysis of review of a book, given a toy corpus data set given below. (Assignment Question)

Document	Category
just plain boring	Negative
entirely predictable and lacks energy	Negative
no surprises and very few laughs	Negative
very powerful book	Positive
the best book of the summer	Positive

## **Course Outcome 5(CO5):**

- 1. Explain lightly supervised approaches to relational analysis.
- 2. Explain a statistical algorithm for word alignment in Machine Translation.

# **Model Question Paper**

QP (	CODE:
Reg Nam	
	Course Code: AMT 302
	Course Name: Concepts in Natural Language Processing
Max	. Marks : 100 Duration: 3 Hours
	PART A
	Answer All Questions. Each Question Carries 3 Marks
1.	Differentiate information extraction and information retrieval.
2.	State Bayes' Theorem.
3.	List threepreprocessing steps that are necessary for an HTML file.
4.	Differentiate CBOW and Skipgram models
5.	Explain the role of support vectors in SVM Classification.
6.	Explain challenges in Name Entity Recognition.
7.	How is a Relational Analysis System evaluated?
8.	Explain the need for an inverted index in an information retrieval system. Are there any more efficient data structures that serve the same purpose.
9.	How do you extract answers to DEFINITION questions?

10.	What are the components that make up a noisy channel model of statistical	
	Machine Translation?	(10x3=30)

		Part B	
	(A	nswer any one question from each module. Each question carries 14 Marks)	
11.	(a)	How is classification done by SVM on linearly separable data?	(8)
	(b)	What is a kernel function? What is the need for a kernel function? Can a kernel function be replaced by an ordinary mapping function?	(4)
	(c)	Explain Heuristic-based NLP.	(2)
		OR	
12.	(a)	Illustrate the steps involved in classification in Naïve Bayes Classifier.	(8)
	(b)	Explain the fundamental tasks that make up an NLP system.	(6)
13.	(a)	Supposing that a set of social media posts' dataset is available to do sentiment analysis. What pre-processing steps need to be done in order to use the data for generating a language model? Illustrate.	(8)
	(b)	Illustrate Bag-of-ngrams model with an example.	(6)
		OR	
14.	(a)	Explain the concept of word embeddings as a model for text representation.	(6)
	(b)	Compare word embeddings model with vectorization approaches.	(4)
	(c)	Explain the concept of feature engineering in NLP Systems.	(4)
15.	(a)	<ol> <li>Given the following data about movie review and its classification, classify"predictable with no fun" to one of the classes using Naïve Bayes Classifier.</li> </ol>	(10)
		Document Category	
		just plain boring Negative	
		entirely predictable and lacks energy Negative	
		no surprises and very few laughs Negative	

very powerful	Positive
the most fun film of the summer	Positive

	(b)	) Explain challenges in Name Entity Recognition.			
		API ABDIORI KALAM			
16.	(a)	Explain Logistic Regression for Text Classification.	(6)		
	(b)	Explain Name Entity Recognition using Sequence Labeling.	(8)		
17.	(a)	Explain supervised approach to relation analysis. What are its limitations?	(10)		
	(b)	How is term selection done for indexing?	(4)		
		OR			
18.	(a)	Given the following data about documents and contents, use tf-idf document scoring method to retrieve the document for the query "sweet love".	(10)		
		Doc 1 Sweet sweet nurse! Love			
		Doc 2 Sweet sorrow			
		Doc 3 How sweet is love?			
		Doc 4 Nurse!			
	(b)	Explain the approaches to evaluate a relation analysis system.	(4)		
19.	(a)	Explain the phases of a factoid question-answering system.	(8)		
	(b)	Give an algorithm for word alignment in Machine Translation.	(6)		
		OR			
20.	(a)	How is decoding done in a Phrase-based Statistical Machine Translation System?	(10)		
	(b)	Explain the concept of Mean Reciprocal Rank.	(4)		

No	Contents	No of Lecture Hrs: 45			
	Module 1 : Introduction to NLP (8 hours)	M			
1.1	Introduction to NLP – Tasks and Applications	1			
1.2	Language – Building Blocks, Challenges of NLP	1			
1.3	Approaches to NLP - Heuristics-Based NLP,	1			
1.4	Machine Learning for NLP	1			
1.5	Naïve Bayes Classifier	1			
1.6	Logistic Regression	1			
1.7	Support Vector Machines – Linearly Separable Data	1			
1.8	Support Vector Machines – Linearly Inseparable Data 1				
	Module 2 : Pre-processing and Representation Models( 10 hour	s)			
2.1	NLP System Pipeline – Stages – Overview	1			
2.2	NLP System Pipeline – Data Acquisition	1			
2.3	NLP System Pipeline – Text Extraction and Cleanup	1			
2.4	NLP System Pipeline – Preprocessing - Sentence segmentation	1			
2.5	Word tokenization, Stemming and lemmatization	1			
2.6	Feature Engineering, Model Building, Evaluation – Metrices, Post- modeling phase	1			
2.7	Text Representation – Vector Space Model	1			
2.8	Vectorization Approaches – One hot encoding, Bag of words	1			

# **TEACHING PLAN**

2.9	Bag of n-grams, TF-IDF	1		
2.10	Word Embeddings – Word2Vec- CBOW, SkipGram models	1		
	Module 3: Classification and Information Extraction(9 hours)			
3.1	Text Classification-Text classification applications, Pipeline for building text classification systems	1		
3.2	Sentiment Analysis using Naïve Bayes Classifier	1		
3.3	Case Studies for Text Classification using Logistic Regression and Support Vector Machines (Lecture 1)	1		
3.4	Case Studies for Text Classification using Logistic Regression and Support Vector Machines (Lecture 2)	1		
3.5	Information Extraction (IE) and Applications, IE Tasks and the IE Pipeline	1		
3.6	Named Entity Recognition (NER), Ambiguity in NER	1		
3.7	NER as Sequence Labeling	1		
3.8	Evaluation of NER	1		
3.9	Practical NER Systems 1			
	Module 4 : Relation Detection and Information Retrieval(8 hour	·s)		
4.1	Relation Detection and Classification – Supervised Learning Approaches to Relation Analysis	1		
4.2	Relation Detection and Classification – Lightly Supervised Approaches to Relation Analysis	1		
4.3	Relation Detection and Classification -Evaluation of Relation Analysis systems	1		
4.4	Information Retrieval – Term weighting and document scoring (Lecture 1)	1		
4.5	Information Retrieval – Term weighting and document scoring (Lecture 2)	1		
4.6	Inverted Index	1		

4.7	Evaluation of Information-Retrieval Systems (Lecture 1)			
4.8	Evaluation of Information-Retrieval Systems (Lecture 2)			
	Module 5 : QA Systems and Machine Translation (10 hours)			
5.1	Question-Answering Systems – Factoid Question Answering, Question Processing	1		
5.2	Passage Retrieval	1		
5.3	Answer Processing, Evaluation of Factoid Answers	1		
5.4	Machine Translation – Why Machine Translation is Hard	1		
5.5	Classical Machine Translation	1		
5.6	Statistical Machine Translation	1		
5.7	The Phrase based Translation model	1		
5.8	Alignment in Machine Translation	1		
5.9	Decoding for Phrase-based Statistical MT (Lecture 1)	1		
5.10	Decoding for Phrase-based Statistical MT (Lecture 2)	1		



AIT304	4 ROBOTICS AND INTELLIGENT SYSTEM	Category	L	Т	Р	Credit	Year of Introduction
		РСС	3	1	0	4 1	2022

## **Preamble**:

This course enables the learners to understand the fundamental concepts and algorithms in Robotics and Intelligent systems. The course covers the standard hardware and kinematic concepts for robot design. Standard algorithms for localization, mapping, path planning, navigation and obstacle avoidance, to incorporate intelligence in robots are included in the course. This course helps the students to design robots with intelligence in a real-world environment.

**Prerequisite:** Basic understanding of probability theory, linear algebra, machine learning, artificial intelligence

## Mapping of course outcomes with program outcomes

CO1	Understand the concepts of manipulator and mobile robotics. (Cognitive Knowledge Level: Understand)
CO2	Choose the suitable sensors, actuators and control for robot design. (Cognitive Knowledge Level: Apply)
CO3	Developing kinematic model of mobile robot and understand robotic vision intelligence. (Cognitive Knowledge Level: Apply)
CO4	Discover the localization and mapping methods in robotics. (Cognitive Knowledge Level: Apply)
CO5	Plan the path and navigation of robot by applying artificial intelligence algorithm. (Cognitive Knowledge Level: Apply)

	PO1	PO2	PO3	PO4	PO5	PO6	<b>PO7</b>	PO8	PO9	PO10	PO11	PO12
C01	$\bigcirc$	F	r'i		$\bigcirc$		0	ĨČ.	ic	- 1	1	$\bigcirc$
CO2	$\bigcirc$			Ń	0	YE!	D'C	ĥ	->	1	-	$\bigcirc$
CO3			0	$\bigcirc$		٢		21 1				$\oslash$
CO4				$\bigcirc$								$\bigcirc$
CO5	Ø			$\bigcirc$								$\oslash$

# Mapping of course outcomes with program outcomes

	Abstract POs defined by National Board of Accreditation							
PO#	Broad PO	P <mark>O</mark> #	Broad PO					
PO1	Engineering Knowledge	PO7	Environment and Sustainability					
PO2	Problem Analysis	PO8	Ethics					
PO3	Design/Development of solutions	PO9	Individual and team work					
PO4	Conduct investigations of complex problems	PO10	Communication					
PO5	Modern tool usage	PO11	Project Management and Finance					
PO6	The Engineer and Society	PO12	Life long learning					

# Assessment Pattern

Bloon 's	1	Continuo	us Assessment Tests	End Semester Examination
Catego ry	A	Test 1 (%)	Test 2 (%)	Marks (%)
Remembe	r	30	30	30
Understan	d	30	30	30
Apply		40	40	40
Analyze				
Evaluate				
Create				

# **Mark Distribution**

Total	CIE	ESE Marks	ESE
Marks	Marks		Duration
150	50	100	3

# **Continuous Internal Evaluation Pattern:**

Attendance	10 marks
Continuous Assessment Tests(Average of Internal Tests 1 & 2)	25 marks
Continuous Assessment Assignment	15 marks

# **Internal Examination Pattern**

Each of the two internal examinations has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed modules and 1 question from the should answer all questions, a student should answer any 5.

### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 full questions from each module of which student should answer any one. Each question can have a maximum 2 subdivisions and carries 14 marks.

# **Syllabus**

### Module – 1 (Introduction to robotics)

Introduction to robotics – Degrees of freedom, Robot types- Manipulators- Anatomy of a robotic manipulator-links, joints, actuators, sensors, controllers. Robot configurations-PPP, RPP, RRP, RRR. Mobile robots- wheeled, legged, aerial robots, underwater robots, surface water robots . Dynamic characteristics- speed of motion, load carrying capacity & speed of response. Introduction to End effectors - mechanical grippers, special tools, Magnetic grippers, Vacuum grippers, adhesive grippers, Active and Passive grippers. Ethics in robotics - 3 laws - applications of robots.

### Module - 2(Sensors, Actuators and Control)

Sensor classification- touch, force, proximity, vision sensors. Internal sensors-Position sensors, velocity sensors, acceleration sensors, Force sensors; External sensors-contact type, non contact type; Digital Camera - CCD camera - CMOS camera - Omnidirectional cameras Sensor characteristics. Actuators - DC Motors - H-Bridge - Pulse Width Modulation - Stepper Motors - Servos, Hydraulic & pneumatic actuators. Control - On-Off Control - PID Control - Velocity Control and Position Control

### Module – 3 (Robotic vision & Kinematics)

Robotic Vision: Sensing, Pre-processing, Segmentation, Description, Recognition, Interpretation, Feature extraction -Camera sensor hardware interfacing. Representation of

Transformations - Representation of a Pure Translation - - Pure Rotation about an Axis - Combined Transformations - Transformations Relative to the Rotating Frame.

Basic understanding of Differential-Drive Wheeled Mobile Robot, Car-Like Wheeled Mobile Robot. Kinematic model of a differential drive and a steered mobile robot, Degree of freedom and manoeuvrability, Degree of steerability, Degree of mobility - different wheel configurations, holonomic and nonholonomic robots. Omnidirectional Wheeled Mobile Robots.

### Module - 4 ( Localization and Mapping)

Position and Orientation - Representing robot position. Basics of reactive navigation; Robot Localization, Challenges in localization - An error model for odometric position estimation

Map Representation - Continuous representations - Decomposition strategies - Current challenges in map representation. Probabilistic map-based localization (only Kalman method), Autonomous map building, Simultaneous localization and mapping (SLAM) - Mathematical definition of SLAM - Visual SLAM with a single camera - Graph-based SLAM - Particle filter SLAM - Open challenges in SLAM

### Module - 5 (Path Planning and Navigation)

Path Planning- Graph search, deterministic graph search - , breadth first search - depth first search - Dijkstra's algorithm, A\*, D\* algorithms, Potential field based path planning. Obstacle avoidance - Bug algorithm - Vector Field Histogram - Dynamic window approaches. Navigation Architectures - Modularity for code reuse and sharing - Control localization - Techniques for decomposition. Alternatives for navigation - Neural networks - Processing the image - Training the neural network for navigation - Convolutional neural network robot control implementation

### **Text Books**

- 1. R Siegwart, IR Nourbakhsh, D Scaramuzza, Introduction to Autonomous Mobile Robots ,, MIT Press, USA, 2011
- 2. Thomas Bräunl Embedded Robotics, Mobile Robot Design and Applications with Embedded Systems-Springer (2006)
- 3. S.G. Tzafestas Introduction to Mobile Robot Control-Elsevier (2014)
- 4. Francis X. Govers Artificial Intelligence for Robotics-Packt Publishing (2018)
- 5. Saeed B. Niku Introduction to Robotics\_ Analysis, Control, Applications

### **Reference Books**

- 1. John J. Craig, Introduction to Robotics, Pearson Education Inc., Asia, 3rd Edition, 2005
- 2. S. K. Saha, Introduction to Robotics 2e, TATA McGraw Hills Education (2014)
- 3. Peter Corke Robotics, Vision and Control\_ Fundamental Algorithms in MATLAB® Springer-Verlag Berlin Heidelberg (2021)

# **Course Level Assessment Questions**

### **Course Outcome1 (CO1):**

- 1. Categorise the various types of Grippers used in robot manipulators.
- 2. Differentiate between active and passive grippers.
- 3. Explain speed of motion and load carrying capacity of a mobile robot.
- 4. You wish to build a dynamically stable robot with a single wheel only. For each of the four basic wheel types, explain whether or not it may be used for such a robot.

### **Course Outcome 2(CO2):**

- 1. Categorise the sensors used in robotics
- 2. Explain any four characteristics of a sensor
- 3. Illustrate the sensor performance measuring parameters
- 4. Suggest any two mechanism to realise 360° Camera

### Course Outcome 3(CO3):

- 1. Determine the degrees of mobility, steerability, and maneuverability for each of the following: (a) bicycle; (b) dynamically balanced robot with a single spherical wheel (c) automobile.
- 2. A frame F was rotated about the y-axis 90°, followed by a rotation about the o-axis of 30°, followed by a translation of 5 units along the n-axis, and finally, a translation of 4 units along the x-axis. Find the total transformation matrix.
- 3. Explain the camera sensor hardware interfacing.
- 4. What is an omni directional robot? Explain two configurations to set up an omni directional robot.

### Course Outcome 4(CO4): .

- 1. Explain the challenges of localization
- 2. How Kalman method can be used in localization of mobile robots
- 3. What are the Decomposition strategies in map representation

4. How Visual SLAM can be performed with a single camera

## **Course Outcome 5(CO5):**

- 1. Explain Dijkstra's algorithm with a suitable example.
- 2. Identify the steps of Generic temporal decomposition of a navigation architecture.
- 3. What is meant by control decomposition? Explain two types of control decomposition.
- 4. Why does SLAM work better with wheel odometer data available?
- 5. In the Floor Finder algorithm, what does the Gaussian blur function does to improve the results?

# **Model Question Paper**

### **QP CODE:**

Reg No: \_\_\_\_\_

Name: \_\_\_\_\_

**PAGES:4** 

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

## SIXTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

**Course Code: AIT 304** 

## **Course Name: ROBOTICS AND INTELLIGENT SYSTEM**

Max. Marks: 100

**Duration: 3** 

### Hours

### PART A

### **Answer All Questions. Each Question Carries 3 Marks**

- 1. What do you mean by degrees of freedom? How many degrees of freedom are required for a drone to achieve any position in 3D space? And how many more DOF required for achieving any orientation as well.
- 2. Explain how leg configuration affects the stability of mobile robot.

- 3. Explain Dynamic range, Linearity and Resolution of a Sensor.
- 4. Explain the working of a Mechanical accelerometer with a block diagram
- 5. Differentiate between holonomic and nonholonomic robots.
- 6. What is the significance of differential drive in mobile robot?
- 7. How will you represent the position and orientation of a wheeled mobile robot?
- 8. Identify the 2 mobile robot localization problems.
- 9. Explain the Bug algorithm for obstacle avoidance.
- 10. What is Voronoi diagram method and its advantages?

(10x3=30)

### Part B

### (Answer any one question from each module. Each question carries 14 Marks)

11.	(a)	Explain the general features of wheeled, legged and aerial robots.	(9)
	(b)	Explain the anatomy of a robotic manipulator with a neat diagram.	(5)
		OR	
12.	(a)	Briefly explain the dynamic characteristics of robots.	(9)
	(b)	Assume an object of mass 140 kg is to be lifted up with an acceleration of 10 m/s2. Calculate the gripper force required for the operation, if coefficient of friction between contact surfaces is 0.2, number of fingers in gripper is 2 and acceleration due to gravity is 9.8 m/s2	(5)
13.	(a)	Explain the working of an Optical Encoder.	(5)
	(b)	A mobile robot is designed for unidirectional motion with constant velocity. Illustrate the mechanism to make the robot move in forward and reverse direction with variable speed. Support with necessary diagrams	(9)

# OR

14.	(a)	Compare and contrast the working of CCD and CMOS camera	(9)
	(b)	Illustrate the significance of the PID controller with a neat block diagram	(5)
15.	(a)	Outline the seven stages of robot vision.	(14)
		OR	
16.	(a)	Derive the kinematic model of a differential drive mobile robot.	(7)
	(b)	A frame B was rotated about the x-axis 90°, then it was translated about the current a-axis 3 inches before it was rotated about the z-axis 90°. Finally, it was translated about the current o-axis 5 inches. (a) Write an equation that describes the motions.	(7)
		(b) Find the final location of a point $p(1,5,4)T$ attached to the frame relative to the reference frame.	
17.	(a)	Derive error model for odometric position estimation	(8)
	(b)	Illustrate the SLAM problem with suitable diagrams	(6)
		OR	
18.	(a)	Compare and Contrast graph based and particle SLAM	(8)
	(b)	Describe the concept of mobile robot localization with suitable Block diagrams	(6)
19.	(a)	Compare and contrast local and global Dynamic window approaches in obstacle avoidance.	(7)
	(b)	Explain the concepts of floor finding Algorithm	(7)

# OR

- 20. (a) Illustrate the Incorporation of Neural network approach in Robot navigation? (6) List its advantages
  - (b) Make the robot to run from start position to goal position in the Following (8) diagram using A\* Algorithm

	1. 1.	1	1.1	1	UNCAL
	Goal		N	VE	RSITY
				Start	
_					

**Teaching Plan** 

No	Estd. Contents	No. of Lecture Hours (45 hrs)	
	<b>Module-1 (Introduction to robotics) (8 hours)</b>		
1.1	Introduction to robotics – Degrees of freedom - Robot types	1 hour	
1.2	Manipulators- Anatomy of a robotic manipulator-links, joints, actuators, sensors, controller	1 hour	

1.3	Robot configurations-PPP, RPP, RRP, RRR- Mobile robots- wheeled	1 hour			
1.4	Legged robots, Aerial robots, underwater robots, surface water robots -	1 hour			
1.5	Dynamic characteristics of robot- speed of motion, load carrying capacity & speed of response	1 hour			
1.6	Introduction to End effectors - mechanical grippers, special tools, Magnetic grippers	1 hour			
1.7	Vacuum grippers, adhesive grippers, Active and Passive grippers	1 hour			
1.8	Ethics in robotics - 3 laws - applications of robots	1 hour			
	Module-2 (Sensors, Actuators and Control) (9 hours)				
2.1	Sensor classification- touch, force, proximity, vision sensors.	1 hour			
2.2	Internal sensors-Position sensors, velocity sensors	1 hour			
2.3	Acceleration sensors, Force sensors;	1 hour			
2.4	External sensors-contact type, non-contact type	1 hour			
2.5	Digital Camera - CCD camera - CMOS camera	1 hour			
2.6	Omnidirectional cameras - Sensor characteristics	1 hour			
2.7	Actuators - DC Motors - H-Bridge - Pulse Width Modulation	1 hour			
2.8	Stepper Motors – Servos - Control - On-Off Control	1 hour			
2.9	PID Control - Velocity Control and Position Control	1 hour			
	Module-3 (Robotic vision & Kinematics) (9 hours)				
3.1	Robot Vision: Sensing, Pre-processing, Segmentation, Description	1 hour			

-

\_

3.2	Recognition, Interpretation, Feature extraction -Camera sensor hardware interfacing	1 hour
3.3	Representation of Transformations - Representation of a Pure Translation - Pure Rotation about an Axis	1 hour
3.4	Combined Transformations - Transformations Relative to the Rotating Frame	1 hour
3.5	Basic understanding of Differential Drive Wheeled Mobile Robot - Car Like Wheeled Mobile Robot	1 hour
3.6	Kinematic model of a differential drive and a steered mobile robot.	1 hour
3.7	Degree of freedom and manoeuvrability, Degree of steerability	1 hour
3.8	Degree of mobility, Different wheel configurations	1 hour
3.9	Holonomic and Nonholonomic robots, Omnidirectional Wheeled Mobile Robots	1 hour
	Module-4 ( Localization and Mapping) (9 hours)	
4.1	Position and Orientation - Representing robot position, Basics of reactive navigation	1 hour
4.2	Robot Localization, Challenges in localization	1 hour
4.3	An error model for odometric position estimation	1 hour
4.4	Map Representation - Continuous representations - Decomposition strategies	1 hour
4.5	Current challenges in map representation, Probabilistic map-based localization (only Kalman method)	1 hour

\_\_\_\_

4.6	Probabilistic map-based localization (only Kalman method)	1 hour
4.7	Autonomous map building, Simultaneous localization and mapping (SLAM) - Mathematical definition of SLAM	1 hour
4.8	Visual SLAM with a single camera - Graph-based SLAM	1 hour
4.9	Particle filter SLAM - Open challenges in SLAM	1 hour
	Module-5 (Path Planning and Navigation) (10 hours)	
5.1	Path Planning- Graph search	1 hour
5.2	Deterministic graph search - breadth first search - depth first search- Dijkstra's algorithm	1 hour
5.3	A*, D* algorithms, Potential field based path planning	1.5 hour
5.4	Obstacle avoidance - Bug algorithm - Vector Field Histogram - Dynamic window approaches	1.5 hour
5.5	Navigation Architectures - Modularity for code reuse and sharing - Control localization - Techniques for decomposition	1 hour
5.6	Alternatives for navigation - Neural networks	1 hour
5.7	Processing the image - Training the neural network for navigation	1.5 hour
5.8	Training the neural network for navigation - Convolutional neural network robot control implementation	1.5 hour

CST	ALGORITHM ANALYSIS AND	Category	L	Т	Р	Credit	Year of Introduction
306	DESIGN	РСС	3	1	0	4	2019

# Preamble:

The course introduces students to the design of computer algorithms, as well as analysis of algorithms. Algorithm design and analysis provide the theoretical backbone of computer science and are a must in the daily work of the successful programmer. The goal of this course is to provide a solid background in the design and analysis of the major classes of algorithms. At the end of the course students will be able to develop their own versions for a given computational task and to compare and contrast their performance.

# **Prerequisite**:

Strong Foundation in Mathematics, Programming in C, Data Structures and Graph Theory.

Course Outcomes: After the completion of the course the student will be able to

CO#	СО
CO1	Analyze any given algorithm and express its time and space complexities in asymptotic notations. (Cognitive Level: Apply)
CO2	Derive recurrence equations and solve it using Iteration, Recurrence Tree, Substitution and Master's Method to compute time complexity of algorithms. (Cognitive Level: Apply)
CO3	Illustrate Graph traversal algorithms & applications and Advanced Datastructures like AVL trees and Disjoint set operations. (Cognitive Level: Apply)
CO4	Demonstrate Divide-and-conquer, Greedy Strategy, Dynamic programming, Branch-and Bound and Backtracking algorithm design techniques (Cognitive Level: Apply)
CO5	Classify a problem as computationally tractable or intractable, and discuss strategies to address intractability (Cognitive Level: Understand)
CO6	Identify the suitable design strategy to solve a given problem. (Cognitive Level: Analyze)

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	٩	٩				T	N			2	
CO2	0	0	0	0	N	2	10	N.	17	- 7	1	
CO3	$\oslash$	0	٢	0		Z	D			20		$\bigotimes$
CO4	Ø	٢	0	0	1 \		1.	10				$\bigcirc$
CO5	Ø											$\checkmark$
CO6	Ø			$\bigcirc$								$\bigcirc$

# Mapping of course outcomes with program outcomes

	Abstract POs defined by	Na <mark>tio</mark> na	l Board of Accreditation
PO#	Broad PO	P <mark>O</mark> #	Broad PO
PO1	Engineering Knowledge	PO7	Environment and Sustainability
PO2	Problem Analysis	PO8	Ethics
PO3	Design/Development of solutions	PO9	Individual and team work
PO4	Conduct investigations of complex problems	PO10	Communication
PO5	Modern tool usage	PO11	Project Management and Finance
PO6	The Engineer and Society	PO12	Life long learning

# **Assessment Pattern**

Bloom's	Continuo	us Assessment Tests	End Semester Examination
Category	Test 1 (%)	Test 2 (%)	Marks (%)
Remember	30	30	30
Understand	30	30	30
Apply	40	40	40

Evaluate	
Create	

**Mark Distribution** 

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3

<b>Continuous Internal Evaluation Pattern:</b>	
Attendance	10 marks
Continuous Assessment Tests (Average of SeriesTests1&2)	25 marks
Continuous Assessment Assignment	15 marks

### **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 questions should answer all questions from the partly completed module), each with 7 marks. Out of the 7 questions, a student should answer any 5.

### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 full questions from each module of which student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.

# Syllabus

# Module-1 (Introduction to Algorithm Analysis)

Characteristics of Algorithms, Criteria for Analysing Algorithms, Time and Space Complexity -Best, Worst and Average Case Complexities, Asymptotic Notations - Big-Oh (O), Big- Omega ( $\Omega$ ), Big-Theta ( $\Theta$ ), Little-oh (o) and Little- Omega ( $\omega$ ) and their properties. Classifying functions by their asymptotic growth rate, Time and Space Complexity Calculation of simple algorithms.

Analysis of Recursive Algorithms: Recurrence Equations, Solving Recurrence Equations – Iteration Method, Recursion Tree Method, Substitution method and Master's Theorem (Proof not required).

## Module-2 (Advanced Data Structures and Graph Algorithms)

Self Balancing Tree - AVL Trees (Insertion and deletion operations with all rotations in detail, algorithms not expected); Disjoint Sets- Disjoint set operations, Union and find algorithms.

DFS and BFS traversals - Analysis, Strongly Connected Components of a Directed graph, Topological Sorting.

# Module-3 (Divide & Conquer and Greedy Strategy)

The Control Abstraction of Divide and Conquer- 2-way Merge sort, Strassen's Algorithm for Matrix Multiplication-Analysis. The Control Abstraction of Greedy Strategy- Fractional Knapsack Problem, Minimum Cost Spanning Tree Computation- Kruskal's Algorithms - Analysis, Single Source Shortest Path Algorithm - Dijkstra's Algorithm-Analysis.

## Module-4 (Dynamic Programming, Back Tracking and Branch & Bound))

The Control Abstraction- The Optimality Principle- Matrix Chain Multiplication-Analysis, All Pairs Shortest Path Algorithm - Floyd-Warshall Algorithm-Analysis. The Control Abstraction of Back Tracking – The N Queen's Problem. Branch and Bound Algorithm for Travelling Salesman Problem.

## **Module-5** (Introduction to Complexity Theory)

Tractable and Intractable Problems, Complexity Classes – P, NP, NP- Hard and NP-Complete Classes- NP Completeness proof of Clique Problem and Vertex Cover Problem- Approximation algorithms- Bin Packing, Graph Coloring. Randomized Algorithms (Definitions of Monte Carlo and Las Vegas algorithms), Randomized version of Quick Sort algorithm with analysis.

## **Text Books**

- 1. T.H.Cormen, C.E.Leiserson, R.L.Rivest, C. Stein, Introduction to Algorithms, 2<sup>nd</sup> Edition, Prentice-Hall India (2001)
- 2. Ellis Horowitz, Sartaj Sahni, Sanguthevar Rajasekaran, "Fundamentals of Computer Algorithms", 2<sup>nd</sup> Edition, Orient Longman Universities Press (2008)

3. Sara Baase and Allen Van Gelder —Computer Algorithms, Introduction to Design and Analysis, 3<sup>rd</sup> Edition, Pearson Education (2009)

### **Reference Books**

- 1. Jon Kleinberg, Eva Tardos, "Algorithm Design", First Edition, Pearson (2005)
- Robert Sedgewick, Kevin Wayne, "Algorithms",4th Edition Pearson (2011)
   Gilles Brassard, Paul Brately, "Fundamentals of Algorithmics", Pearson (1996)
- Steven S. Skiena, "The Algorithm Design Manual", 2<sup>nd</sup> Edition, Springer(2008)

# **Course Level Assessment Questions**

## **Course Outcome 1 (CO1):**

- 1. Is  $2^{n+1} = O(2^n)$ ? Is  $2^{2n} = O(2^n)$ ? Justify your answer.
- 2. What is the need of asymptotic analysis in calculating time complexity? What are the notations

used for asymptotic analysis?

- 3. Calculate the time complexity for addition of two matrices.
- 4. Define time complexity and space complexity. Write an algorithm for adding n natural numbers and analyse the time and space requirements of the algorithm.

## Course Outcome 2 (CO2):

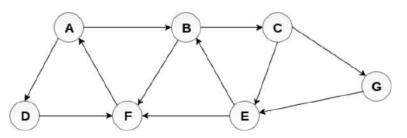
- 1. State Master's theorem for solving recurrences.
- 2. Solve the recurrence T(n) = 3T(n-2), using iteration method
- 3. State the conditions in recurrences where Master Theorem is not applicable.
- 4. Solve the following recurrence equations using Master's theorem.

a) T (n) = 
$$8T(n/2) + 100 n^2$$
  
b) T (n) =  $2T(n/2) + 10 n$ 

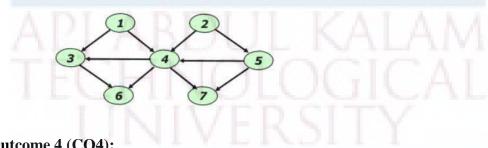
5. Using Recursion Tree method, Solve T(n) = 2T(n/10) + T(9n/10) + n. Assume constant time for small values of n.

## Course Outcome 3 (CO3):

- 1. Explain the rotations performed for insertion in AVL tree with example.
- 2. Write down BFS algorithm and analyse the time complexity. Perform BFS traversal on the given graph starting from node A. If multiple node choices are available for next travel, choose the next node in alphabetical order.

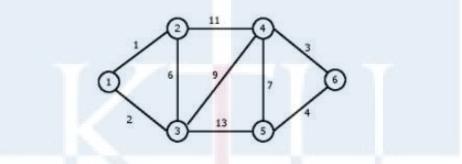


- 3. Find the minimum and maximum height of any AVL-tree with 7 nodes? Assume that the height of a tree with a single node is 0. (3)
- 4. Find any three topological orderings of the given graph.

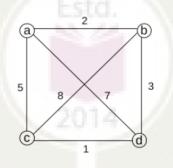


# Course Outcome 4 (CO4):

- 1. Give the control abstraction for Divide and Conquer method.
- 2. Construct the minimum spanning tree for the given graph using Kruskal's algorithm. Analyse the complexity of the algorithm.



- 3. Compare Divide and Conquer and Dynamic programming methodologies
- 4. What is Principle of Optimality?
- 5. Define Travelling Salesman Problem (TSP). Apply branch and bound algorithm to solve TSP for the following graph, assuming the start city as 'a'. Draw the state space tree.



## **Course Outcome 5 (CO5):**

- 1. Compare Tractable and Intractable Problems
- 2. With the help of suitable code sequence convince Vertex Cover Problem is an example of NP-Complete Problem

- 3. Explain Vertex Cover problem using an example. Suggest an algorithm for finding Vertex Cover of a graph.
- 4. Write short notes on approximation algorithms.
- 5. Compare Conventional quick sort algorithm and Randomized quicksort with the help of a suitable example?

Course Outcome 6 (CO6): (CO attainment through assignment only, not meant for examinations)

Choosing the best algorithm design strategy for a given problem after applying applicable design strategies – Sample Problems Given.

- 1. Finding the Smallest and Largest elements in an array of 'n' numbers
- 2. Fibonacci Sequence Generation.
- 3. Merge Sort
- 4. Travelling Sales Man Problem
- 5. 0/1 Knapsack Problem

# **Model Question Paper**

# **QP CODE:**

Reg No: \_\_\_\_\_

Name: \_\_\_\_\_

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY PAGES : 4

## SIXTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

Course Code: CST 306

**Course Name: Algorithm Analysis and Design** 

Max. Marks: 100

**Duration: 3 Hours** 

# PART A

# Answer All Questions. Each Question Carries 3 Marks

1. Define asymptotic notation? Arrange the following functions in increasing order of asymptotic growth rate.

n³, 2<sup>n</sup>, log n³, 2<sup>100</sup>, n² log n, n<sup>n</sup>, log n, n<sup>0.3</sup>, 2<sup>logn</sup>

- 2. State Master's Theorem. Find the solution to the following recurrence equations using Master's theorem.
  - a) T (n) =  $8T(n/2) + 100 n^2$ b) T (n) = 2T(n/2) + 10 n
- 3. Find any two topological ordering of the DAG given below.
- 4. Show the UNION operation using linked list representation of disjoint sets.
- 5. Write the control abstraction of greedy strategy to solve a problem.
- 6. Write an algorithm based on divide-and-conquer strategy to search an element in a given list. Assume that the elements of list are in sorted order.
- 7. List the sequence of steps to be followed in Dynamic Programming approach.
- 8. Illustrate how optimal substructure property could be maintained in Floyd-Warshall algorithm.
- 9. Differentiate between P and NP problems.
- 10. Specify the relevance of approximation algorithms.

(10x3=30)

## Part B

## (Answer any one question from each module. Each question carries 14 Marks)

- 11. (a) Define Big O, Big  $\Omega$  and Big  $\Theta$  Notation and illustrate them graphically. (7)
  - (b) Solve the following recurrence equation using recursion tree method T(n) = T(n/3) + T(2n/3) + n, where n>1 T(n) = 1, Otherwise
    (7)

(7)

(7)

(7)

- Explain the iteration method for solving recurrences and solve the following 12. (a) (7) recurrence equation using iteration method. T(n) = 3T(n/3) + n; T(1) = 1
  - Determine the time complexities of the following two functions fun1() and (b) fun2().

int fun1(int n) if  $(n \le 1)$  return n; return 2\*fun1(n-1); int fun2 (int n)

i)

ii)

if  $(n \le 1)$  return n; return fun2 (n-1) + fun2 (n-1)}

Write DFS algorithm and analyse its time complexity. Illustrate the 13. (a) classification of edges in DFS traversal.

ł

Find the strongly connected components of the digraph given below: (b)

В D Н A С

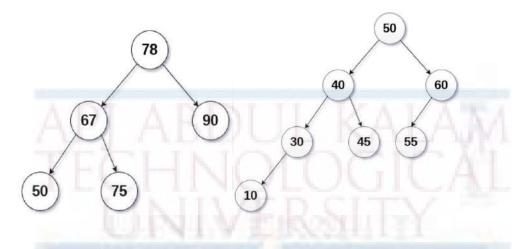
Illustrate the advantage of height balanced binary search trees over binary 14. (a) (7) search trees? Explain various rotations in AVL trees with example.

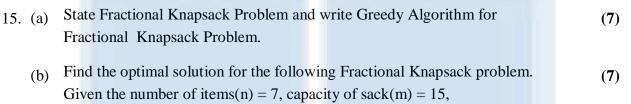
Perform the following operations in the given AVL trees. (b) (7)

i) Insert 70

ii) Delete 55



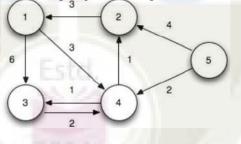




 $W = \{2,3,5,7,1,4,1\}$  and  $P = \{10,5,15,7,6,18,3\}$ 

### OR

- 16. (a) Write and explain merge sort algorithm using divide and conquer strategy (7) using the data {30, 19, 35, 3, 9, 46, 10}. Also analyse the time complexity.
  - (b) Write the pseudo code for Dijkstra's algorithm. Compute the shortest distance (7) from vertex 1 to all other vertices using Dijkstra's algorithm.



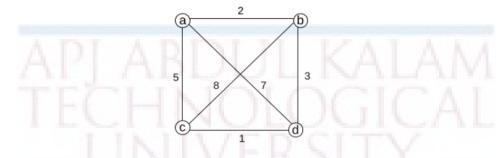
(5)

- 17. (a) Write Floyd-Warshall algorithm and analyse its complexity.
  - (b) Write and explain the algorithm to find the optimal parenthesization of matrix (9) chain product whose sequence of dimension is 4x10,10x3, 3x12,12x20.

### OR

18. (a) Explain the concept of Backtracking method using 4 Queens problem. (7)

(b) Define Travelling Salesman Problem (TSP). Apply branch and bound (7) algorithm to solve TSP for the following graph, assuming the start city as 'a'. Draw the state space tree.



State bin packing problem? Explain the first fit decreasing strategy	(7)
Prove that the Clique problem is NP-Complete.	(7)
OR	
Explain the need for randomized algorithms. Differentiate Las Vegas and Monte Carlo algorithms	(6)
	Prove that the Clique problem is NP-Complete.

(b) Explain randomized quicksort and analyse the expected running time of randomized quicksort with the help of a suitable example? (9)

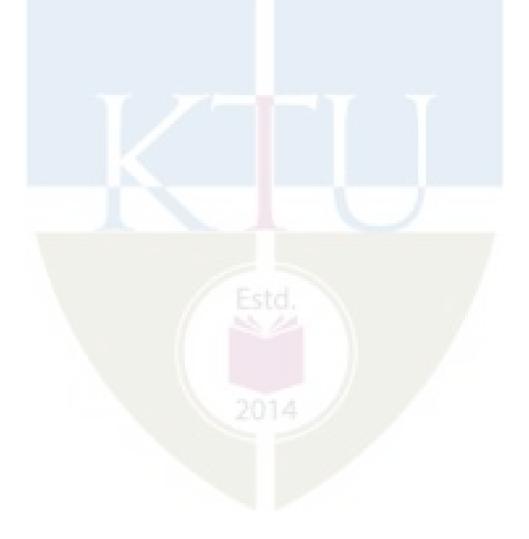
# **Teaching Plan**

No	Торіс	No. of Hours (45 hrs)
	Module -1 (Introduction to Algorithm Analysis) 9 hrs.	
1.1	Introduction to Algorithm Analysis: Characteristics of Algorithms.	1 hour
1.2	Criteria for Analysing Algorithms, Time and Space Complexity - Best, Worst and Average Case Complexities.	1 hour
1.3	Asymptotic Notations - Properties of Big-Oh (O), Big-Omega ( $\Omega$ ), Big-Theta ( $\Theta$ ), Little-Oh (o) and Little-Omega ( $\omega$ ).	1 hour
1.4	Illustration of Asymptotic Notations	1 hour

1.5	Classifying functions by their asymptotic growth rate	1 hour
1.6	Time and Space Complexity Calculation of algorithms/code segments.	1 hour
1.7	Analysis of Recursive Algorithms: Recurrence Equations, Solving Recurrence Equations – Iteration Method.	1 hour
1.8	Recursion Tree Method	1 hour
1.9	Substitution method and Master's Theorem and its Illustration.	1 hour
	Module-2 (Advanced Data Structures and Graph Algorithms) 10	Hrs.
2.1	Self Balancing Trees - Properties of AVL Trees, Rotations of AVL Trees	1 hour
2.2	AVL Trees Insertion and Illustration	1 hour
2.3	AVL Trees Deletion and Illustration	1 hour
2.4	Disjoint set operations.	1 hour
2.5	Union and find algorithms.	1 hour
2.6	Illustration of Union and find algorithms	1 hour
2.7	Graph Algorithms: BFS traversal, Analysis.	1 hour
2.8	DFS traversal, Analysis.	1 hour
2.9	Strongly connected components of a Directed graph.	1 hour
2.10	Topological Sorting.	1 hour
	Module-3 (Divide & Conquer and Greedy Method) 8 Hrs	
3.1	Divide and Conquer: The Control Abstraction.	1 hour
3.2	2-way Merge Sort, Analysis.	1 hour
3.3	Strassen's Algorithm for Matrix Multiplication, Analysis	1 hour

3.4	Greedy Strategy: The Control Abstraction.	1 hour
3.5	Fractional Knapsack Problem.	1 hour
3.6	Minimum Cost Spanning Tree Computation- Kruskal's Algorithm, Analysis.	1 hour
3.7	Single Source Shortest Path Algorithm - Dijkstra's Algorithm	1 hour
3.8	Illustration of Dijkstra's Algorithm-Analysis.	1 hour
	Module-4 (Dynamic Programming, Back Tracking and Branch and H	Bound) 8 Hrs.
4.1	Dynamic Programming: The Control Abstraction, The Optimality Principle.	1 hour
4.2	Matrix Chain Multiplication-Analysis.	1 hour
4.3	Illustration of Matrix Chain Multiplication-Analysis.	1 hour
4.4	All Pairs Shortest Path Algorithm- Analysis and Illustration of Floyd-Warshall Algorithm.	1 hour
4.5	Back Tracking: The Control Abstraction .	1 hour
4.6	Back Tracking: The Control Abstraction – The N Queen's Problem.	1 hour
4.7	Branch and Bound:- Travelling salesman problem.	1 hour
4.8	Branch and Bound:- Travelling salesman problem.	1 hour
	Module-5 (Introduction to Complexity Theory) 10 Hrs	
5.1	Introduction to Complexity Theory: Tractable and Intractable Problems.	1 hour
5.2	Complexity Classes – P, NP.	1 hour
5.3	NP- Hard and NP-Complete Problems.	1 hour
5.4	NP Completeness Proof of Clique Problem.	1 hour

5.5	NP Completeness Proof of Vertex Cover Problem.	1 hour	
5.6	Approximation algorithms- Bin Packing Algorithm and Illustration.	1 hour	
5.7	Graph Colouring Algorithm and Illustration.	1 hour	
5.8	Randomized Algorithms (definitions of Monte Carlo and Las Vegas algorithms).	1 hour	
5.9	Randomized Version of Quick Sort Algorithm with Analysis.	1 hour	
5.10	Illustration of Randomized Version of Quick Sort Algorithm with Analysis.	1 hour	



# API ABDULKALAM SEMESTER VI PROGRAM ELECTIVE I



AIT 312	RECOMMENDATI ON SYSTEM	CATEGORY	L	Т	Р	CREDIT	YEAR OF INTRODUCTION
		PEC	2	1	0	3	2020

**Preamble:** The course is prepared with the view of facilitating the learner to get an overview of recommender system. This course covers the concepts like Introduction to basic concepts and Recent developments, Collaborative Filtering, Content-based recommendation, Knowledge based recommendation, Hybrid approaches and Evaluating Recommender System. The course enables the learners to develop state-of-the-art recommender systems that automate a variety of choice-making strategies with the goal of providing affordable, personal, and high-quality recommendations

### Prerequisite: Nil

**Course Outcomes:** After the completion of the course the student will be able to:

CO 1	Describe the basic concepts of recommender systems (Cognitive Knowledge Level : Understand)									
CO 2	Summarize the features of constraint based and case-based knowledge- based recommender systems (Cognitive Knowledge Level : Understand)									
CO 3	Explain different hybridizing algorithms and illustrate them with suitable examples. (Cognitive Knowledge Level : Understand )									
CO 4	Analyze the design issues in offline recommender evaluation (Cognitive Knowledge Level : Apply)									
CO 5	Explain the features of attack-resistant recommender systems (Cognitive Knowledge Level : Understand)									

### Mapping of course outcomes with program outcomes

	PO 1	PO 2	PO 3	<b>PO</b> 4	<b>PO</b> 5	PO6	PO 7	PO 8	PO 9	PO1 0	PO1 1	PO1 2
CO1						2	014	2	/			
CO2	$\bigotimes$	$\bigcirc$	$\bigotimes$									
CO3	$\bigcirc$	$\bigcirc$										
CO4	$\bigcirc$	$\oslash$	$\bigcirc$									

CO5					
	Abstract POs defined by Nation	al Boar	d of Accreditation		
PO#	Broad PO	PO#	Broad PO		
PO1	Engineering Knowledge	PO7	Environment and Sustainability		
PO2	Problem Analysis PO8 Ethics		Ethics		
PO3	Design/Development of solutions	ign/Development of solutions <b>PO9</b> Individual and team work			
PO4	Conduct investigations of complex problems	PO10	0 Communication		
PO5	Modern tool usage	PO11	Project Management and Finance		
PO6	The Engineer and Society	PO12	Lifelong learning		

# **Assessment Pattern**

	Continuous Asses	sment Tests		6
Bloom's Category	Test1 (percentage)	Test2 (percentage)	– End Examinatio	Semester n Marks
Remember	40	40	40	
Understand	40	40	40	
Apply	20	20	20	
Analyze				
Evaluate	1			
Create		std.		

#### **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3

# **Continuous Internal Evaluation Pattern:**

#### Attendance

Continuous Assessment Tests(Average of Internal Tests 1 & 2)

10 marks 25 marks

#### Continuous Assessment Assignment

#### 15 marks

#### **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks.First Internal Examination shall be preferably conducted after completing the first half of the syllabus and the Second Internal Examination shall be preferably conducted after completing the remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly covered module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 questions (preferably, 3 each with 7 marks. Out of the 7 questions in Part B, a student should answer any 5.

#### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have a maximum of 2 subdivisions and carries 14 marks.

#### **Syllabus**

#### Module 1(Introduction to basic concepts and Recent developments)

Introduction to basic concepts and Recent developments, Collaborative recommendation -User-based nearest neighbor recommendations, Item-based nearest neighbour recommendation, Collaborative recommendation ratings, Model-based and preprocessingbased approaches, Recent practical approaches and systems Content-based recommendation -Content representation and content similarity Similarity-based retrieval and other text classification methods

#### Module 2 (Knowledge-based recommendation)

Knowledge-based recommendation - Knowledge representation and reasoning, Constraints, Cases and similarities, Interacting with constraint-based recommenders - Defaults Dealing with unsatisfiable requirements and empty result set, Proposing repairs for unsatisfiable requirements, Ranking the items/utility-based recommendation, Interacting with case-based recommenders, Critiquing -Compound critiquing, Dynamic critiquing

#### Module 3 (Hybrid recommendation approaches )

Hybrid recommendation approaches - Opportunities for hybridization Recommendation paradigms, Hybridization designs, Monolithic hybridization design - Feature

combination hybrids, Feature augmentation hybrids, Parallelized hybridization design -Mixed hybrids, Switching hybrids, Weighted hybrids, Pipelined hybridization design Cascade hybrids, Meta-level hybrids Limitations of hybridization strategies

#### Module 4 (Evaluating Recommender Systems)

Introduction - Evaluation Paradigms, User Studies, Online Evaluation Offline Evaluation with Historical Data Sets, General Goals of Evaluation Design - Accuracy, Coverage, Confidence and Trust, Novelty, General Goals of Evaluation Design - Serendipity, Diversity, Robustness and Stability Scalability, Design Issues in Offline Recommender Evaluation - Case Study of the Netflix Prize Data Set, Segmenting the Ratings for Training and Testing - Hold-Out, Cross-Validation, Comparison with Classification, Accuracy Metrics in Offline Evaluation - Measuring the Accuracy of Ratings Prediction, RMSE versus MAE, Impact of the Long Tail, Evaluating Ranking via Correlation, Evaluating Ranking via Utility Evaluating Ranking via Receiver Operating Characteristic, Limitations of Evaluation Measures - Avoiding Evaluation Gaming

#### Module 5 (Attack-Resistant Recommender Systems)

Introduction Understanding the Trade-Offs in Attack Models - Quantifying Attack Impact Types of Attacks - Random Attack . Average Attack Bandwagon, Reverse Bandwagon Attack, Probe Attack Segment Attack, Effect of Base Recommendation Algorithm, Detecting Attacks on Recommender Systems - Individual Attack Profile Detection , Group Attack Profile Detection - Preprocessing Methods Online Methods Strategies for Robust Recommender Design - Preventing Automated Attacks with CAPTCHAs Using Social Trust. Designing Robust Recommendation Algorithms - Incorporating Clustering in Neighborhood Methods Fake Profile Detection during Recommendation Time Association - Based Algorithms

#### **Text Books**

- 3. Jannach D., Zanker M. and FelFering A., Recommender Systems: An Introduction, Cambridge University Press(2011)
- 4. C.C. Aggarwal, Recommender Systems: The Textbook, Springer, 2016.

#### **Reference Books**

- 5. F. Ricci, L Rokach, B. Shapira and P.B. Kantor, Recommender systems handbook, Springer 2010
- Manouselis N., Drachsler H., Verbert K., Duval E., Recommender Systems For Learning, Springer (2013), 1st ed.

# **Course Level Assessment Questions**

**Course Outcome 1 (CO1):** Discuss the cases in which content-based recommendations will not perform as well as collaborative filtering.

**Course Outcome 2 (CO2):** Analyze, in detail, different techniques available to support users in the interaction with constraint-based recommender applications.

**Course Outcome 3(CO3):** Explain about the feature combination and feature augmentation hybrid mechanisms.

**Course Outcome 4 (CO4):** With appropriate case study, explain the design issues in offline recommender evaluation .

**Course Outcome 5 (CO5):** Illustrate different methods used to detect attacks on existing recommender system.

# **Model Question Paper**

#### **QP CODE:**

Reg No: \_\_\_\_\_

Name: \_\_\_\_\_

**PAGES:3** 

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

#### SIXTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

#### Course Code: AIT 312

**Course Name: Recommendation System** 

Max. Marks : 100

**Duration: 3 Hours** 

# PART A

#### **Answer All Questions. Each Question Carries 3 Marks**

- 4. List any two purposes of recommender systems.
- 5. Indicate the main idea of collaborative recommendation approaches.
- 6. Describe the two types of outputs generated with pure collaborative approaches that takes matrix of given user-item ratings as the only input .
- 7. Define case amplification. How it can be computed?
- 8. Explain the need of item-based nearest neighbor recommendation system.
- 9. Define classical constraint satisfaction problem (CSP).

(10x3=30)

- 10. Differentiate between parallelized hybridization design and pipelined hybridization design with suitable diagram.
- 11. What is meant by monolithic hybridization design?
- 12. Differentiate between internal validity and external validity.
- 13. Specify the effect of base recommendation algorithm.

#### Part B

(Answer any one question from each module. Each question carries 14 Marks)

14.	(a)	Describe about user-based nearest neighbor recommendation system which deals with new items for which no ratings exist.	(7)
	(b)	Explain about Rocchio's relevance feedback method	(7)
		OR	
15.	(a)	Summarize the implicit and explicit rating mechanism in collaborative recommendation approaches.	(7)

- (b) Explain any two techniques that deal with data sparsity and the cold-start (7) problem.
- 16. (a) Explain *QuickXPlain* algorithm that calculates one conflict set at a time for (7) a given set of constraints.
  - (b) Which are the ways available to specify defaults? Explain how derived (7) defaults can be determined.

#### OR

17.	(a)	Explain about the ranking of items/utility-based recommendation.	(7)
	(b)	Describe <i>DynamicCritiquing</i> algorithm.	(7)
18.	(a)	Explain about feature combination hybrids.	(7)
	(b)	Describe feature augmentation hybrid.	(7)

OR

19. (a) Explain about different parallelized hybridization strategies. (7)

LEARNING

	(b)	Describe pipelined hybridization design methods.	(7)
20.	(a)	Explain about offline and online evaluations in recommender systems.	(7)
	(b)	Describe the general goals of evaluation design.	(7)
		OR	
21.	(a)	Discuss about the design issues in offline recommender evaluation. Illustrate with a case study.	(7)
	(b)	Explain about accuracy metrics in offline evaluation.	(7)
22.	(a)	How do you quantify attack impact on recommender system?	(7)
	(b)	Discuss about different attacks on recommender system.	(7)
		OR	
23.	(a)	Discuss about different methods available to detect attacks on recommender system.	(7)

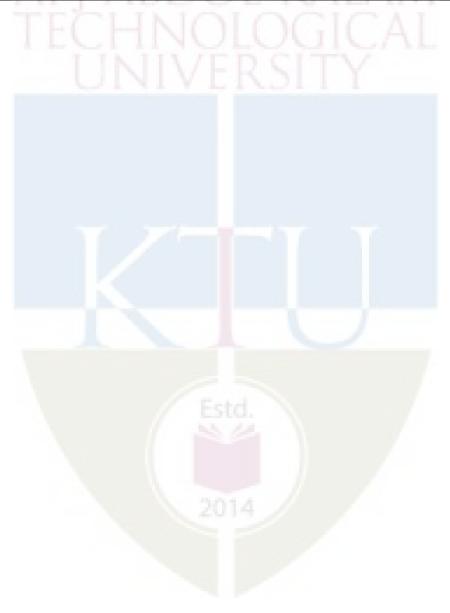
(b) Explain how to design robust recommendation algorithms.

(7)

	Teaching Plan	
No	Contents	No. of Lecture Hours ( 36 hrs)
Mo	dule - 1 (Introduction to basic concepts )	(7 hours)
1.1	Introduction to basic concepts and Recent developments	1 hour
1.2	Collaborative recommendation :User-based nearest neighbor recommendation	1 hour
1.3	Collaborative recommendation :Item-based nearest neighbor recommendation	1 hour
1.4	Collaborative recommendation ratings	1 hour
1.5	Model-based and preprocessing-based approaches, Recent practical approaches and systems	1 hour
1.6	Content-based recommendation - Content representation and content similarity	1 hour
1.7	Similarity-based retrieval and Other text classification methods	1 hour
Mo	odule - 2 (Knowledge-based recommendation)	(8 hours)

2.1 Knowledge representation and reasoning , Constraints, Cases and similarities	1 hour
2.2 Interacting with constraint-based recommenders - Defaults	1 hour
2.3 Dealing with unsatisfiable requirements and empty result sets	1 hour
2.4 Proposing repairs for unsatisfiable requirements	1 hour
2.5 Ranking the items/utility-based recommendation	1 hour
2.6 Interacting with case-based recommenders Introduction	1 hour
2.7 Critiquing -Compound critiquing	1 hour
2.8 Dynamic critiquing	1 hour
Module- 3 (Hybrid recommendation)	(7 hours)
3.1 Hybrid recommendation approaches - Opportunities for hybridization Recommendation paradigms, Hybridization designs	1 hour
3.2 Monolithic hybridization design - Feature combination hybrids	1 hour
3.3 Feature augmentation hybrids	1 hour
3.4 Parallelized hybridization design -Mixed hybrids, Switching hybrids	1 hour
3.5 Weighted hybrids	1 hour
3.6 Pipelined hybridization design Cascade hybrids, Meta-level hybrids	1 hour
3.7 Limitations of hybridization strategies	1 hour
Module - 4 ( Evaluating Recommender Systems )	(8 hours)
4.1 Introduction - Evaluation Paradigms , User Studies , Online Evaluation Offline Evaluation with Historical Data Sets	1 hour
4.2 Goals of Evaluation Design - Accuracy, Coverage, Confidence and Trust, Novelty	1 hour
4.3 General Goals of Evaluation Design - Serendipity ,Diversity , Robustness and Stability Scalability	1 hour
4.4 Design Issues in Offline Recommender Evaluation - Case Study of the Netflix Prize Data Set	1 hour
4.5 Design Issues in Offline Recommender Evaluation -Segmenting the Ratings for Training and Testing - Hold-Out, Cross-Validation, Comparison with Classification	
4.6 Accuracy Metrics in Offline Evaluation - Measuring the Accuracy of Ratings Prediction, RMSE versus MAE, Impact of the Long Tail	1 HOUI
4.7 Evaluating Ranking via Correlation , Evaluating Ranking via Utility Evaluating Ranking via Receiver Operating Characteristic	
4.8 Limitations of Evaluation Measures - Avoiding Evaluation Gaming	1 hour
Module- 5 (Attack-Resistant Recommender Systems )	(6 hours)
5.1 Introduction Understanding the Trade-Offs in Attack Models - Quantifying Attack Impact	1 hour
5.2 Types of Attacks - Random Attack . Average Attack Bandwagon	1 hour
5.3 Reverse Bandwagon Attack, Probe Attack Segment Attack, Effect of Base	1 hour

	Recommendation Algorithm	
5.4	Detecting Attacks on Recommender Systems - Individual Attack Profile Detection ,Group Attack Profile Detection - Preprocessing Methods Online Methods	1 hour
5.5	Strategies for Robust Recommender Design - Preventing Automated Attacks with CAPTCHAs Using Social Trust . Designing Robust Recommendation Algorithms - Incorporating Clustering in Neighborhood Methods Fake Profile Detection during Recommendation Time	1 hour
5.6	Association-Based Algorithms	1 hour



AIT	CONCEPTS IN COMPUTER	Category	L	Т	Р	Credit	Year of Introduction
322	GRAPHICS AND IMAGE PROCESSING	PEC	2	1	0	4	2019

#### **Preamble:**

The purpose of this course is to make awareness about strong theoretical relationships between computer graphics and image processing. This course helps the learner to understand three-dimensional environment representation in a computer, transformation of 2D/3D objects, basic mathematical techniques and algorithms used to build useful applications, imaging, and image processing techniques. The study of computer graphics and image processing develops the abilityto create image processing frameworks for different domains and develops algorithms for emerging display technologies.

**Prerequisite:** A sound knowledge of Mathematics and a programming language.

Course Outcomes: After the completion of the course the student will be able to

CO#	СО
C01	Describe the working principles of graphics devices(Cognitive Knowledge level: Understand)
CO2	Illustrate line drawing, circle drawing and polygon filling algorithms(Cognitive Knowledge level: Apply)
CO3	Demonstrate geometric representations, transformations on 2D & 3D objects, clipping algorithms and projection algorithms(Cognitive Knowledge level: Apply)
CO4	Summarize visible surface detection methods(Cognitive Knowledge level: Understand)
CO5	Summarizethe concepts of digital image representation, processing and demonstrate pixel relationships(Cognitive Knowledge level: Apply)
CO6	Solve image enhancement and segmentation problems using spatial domain techniques(Cognitive Knowledge level: Apply)

LEARNING
Mapping of course outcomes with program outcomes

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO1 0	PO11	PO1 2
CO1	$\bigcirc$		T	A D		TT	V	1 A 1	A			
CO2	Ø	۲	0	0	닢	51	2		4	V1		
CO3	$\oslash$	0	0	0	Ň	22	Ň	11	ĻΡ	L.		
CO4	Ø		0	N.	V	ΕK	S					
CO5	0	$\bigcirc$										
CO6	Ø					0						

		Abstract POs defined b	y Nation	al Board of Accreditation
PO#		Broad PO	PO#	Broad PO
PO1	Engine	eering Knowledge	PO7	Environment and Sustainability
PO2	Proble	m Analysis	PO8	Ethics
PO3	Design	n/Development of solutions	PO9	Individual and team work
PO4	Condu proble	ct investigations of complex ms	PO10	Communication
PO5	Moder	n tool usage	PO11	Project Management and Finance
PO6	The E	ngineer and Society	PO12	Life long learning

#### **Assessment Pattern**

Bloom's	Continuou	s Assessment Tests	End Semester Examination
Category	Test 1 (%)	Test 2 (%)	Marks (%)
Remember	30	30	30
Understand	30	30	30
Apply	40	40	40
Analyze			

#### LEARNING

Evaluate		
Create		

#### **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3

# **Continuous Internal Evaluation Pattern:**

Attendance		10 marks
Continuous As	sessment Tests(Average of SeriesTests1&2)	25 marks
<i>.</i>		

Continuous Assessment Assignment15 marks

#### **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. The first series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing the remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed module), each with 7 marks. Out of the 7 questions, a student should answer any5.

#### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 full questions from each module of which student should answer any one full question. Each question can have maximum 2 sub-divisions and carries 14 marks.

# **Syllabus**

#### Module – 1(Basics of Computer graphics and Algorithms)

Basics of Computer Graphics and its applications. Video Display devices- Refresh Cathode Ray Tubes, Random Scan Displays and systems, Raster scan displays and systems. Line

B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING drawingalgorithms- DDA, Bresenham's algorithm. Circle drawing algorithms- Midpoint Circle generation algorithm, Bresenham's algorithm.

# Module - 2(Filled Area Primitives and transformations)

Filled Area Primitives- Scan line polygon filling, Boundary filling and flood filling. Two dimensional transformations-Translation, Rotation, Scaling, Reflection and Shearing, Composite transformations, Matrix representations and homogeneous coordinates. Basic 3D transformations.

# Module - 3 (Clipping and Projections)

Window to viewport transformation. Cohen Sutherland Line clipping algorithm.Sutherland Hodgeman Polygon clipping algorithm. Three-dimensional viewing pipeline. Projections-Parallel and Perspective projections. Visible surface detection algorithms- Depth buffer algorithm, Scan line algorithm.

# Module - 4 (Fundamentals of Digital Image Processing)

Introduction to Image processing and applications. Image as 2D data. Image representation ingrayscale, Binary and Colour images. Fundamental steps in image processing. Components of image processing system.Coordinate conventions. Sampling and quantization. Spatial and Gray Level Resolution. Basic relationship between pixels– neighbourhood, adjacency, connectivity.

# Module - 5 (Image Enhancement in Spatial Domain and Image Segmentation)

Basic gray level transformation functions- Log transformations, Power-Law transformations, Contrast stretching. Histogram equalization. Basics of spatial filtering - Smoothing spatial filter-Linear and nonlinear filters, and Sharpening spatial filters-Gradient and Laplacian.

Fundamentals of Image Segmentation. Thresholding-Basics of Intensity thresholding and Global Thresholding. Region based Approach- Region Growing, Region Splitting and Merging. Edge Detection - Edge Operators- Sobel and Prewitt.

# **Text Book**

- 5. Donald Hearn and M. Pauline Baker, Computer Graphics, PHI, 2e, 1996
- 6. Rafael C. Gonzalez and Richard E. Woods, Digital Image Processing. Pearson, 4e, 2017

# References

- a. William M. Newman and Robert F. Sproull, Principles of Interactive Computer Graphics. McGraw Hill, 2001
- b. Zhigang Xiang and Roy Plastock, Computer Graphics (Schaum's outline Series), McGraw Hill, 2019.
- c. David F. Rogers , Procedural Elements for Computer Graphics, Tata McGraw Hill, 2001.
- d. M. Sonka, V. Hlavac, and R. Boyle, Image Processing, Analysis, and Machine Vision, Thomson India Edition, 4e, 2017.

# **Course Level Assessment Questions**

#### **Course Outcome 1 (CO1):**

- 4. Compare the working principle of raster scan systems and random scan systems.
- 5. How much time is spent scanning across each row of pixels during screen refresh on a raster system with resolution of 1280\*1024 and a refresh rate of 60 frames per second?

#### Course Outcome 2 (CO2):

- 3. Rasterize the line with end points(2,3) and (5,8) using Bresenham's line drawing algorithm.
- 4. Explain how the 4-connected area filling approach differs from 8- connected area filling in boundary filling algorithm

#### Course Outcome 3 (CO3):

**21.** Rotate a triangle ABC 45 degree counter clockwise about the pivot point (10,3), where the position vector of the coordinate ABC is given as A(4,1), B(5,2) and C(4,3).

**22.** Given a clipping window A(20,20), B(60,20), C(60,40) and D(20,40). Using Cohen Sutherland algorithm, find the visible portion of the line segment joining the points P(40,80) and Q(120,30)

#### **Course Outcome 4 (CO4):**

2. Explain scan line algorithm for detecting visible surfaces in an object.

#### **Course Outcome 5 (CO5):**

- 1. Give an image representation model and describe how the representation changes in grayscale, binary and colour images.
- 2. Consider an image segment shown below.
  - 3 1 2 1 (q)
  - $2\ 2\ 0\ 2$
  - 1 2 1 1
    - (p) 1 0 1 2
  - (a) Let  $V = \{0,1\}$  and compute the length of the shortest 4-,8- and m- path between p and

q. If a particular path does not exist between these two points, explain why?

- (b) Repeat for  $V = \{1, 2\}$ .
- 3. The spatial resolution of an image is given by 128 X 128. What is its storage requirements if it is represented by 64 gray levels?

#### **Course Outcome 6 (CO6):**

6. A skilled medical technician is charged with the job of inspecting a certain class of monochrome images generated by electronic microscope. To facilitate the inspection, the technician uses image

processing aids. However when he examines the images he finds the following problems.

- 4. Presence of bright isolated dots that are not of interest.
- 5. Lack of sharpness
- 6. Poor contrast

Identify the sequence of preprocessing steps that the technician may use to overcome the above-mentioned problems and explain it.

7. A 4x4, 4 bits/pixel original image is given by

- 5. Apply histogram equalisation to the image by rounding the resulting image pixels to integers
- 6. Sketch the histogram of the original image and the histogram-equalised image.
- 8. You have Sobel operator and Laplacian operator for edge detection. Which operator will you select for edge detection in the case of noisy image? Explain.(Assignment)

# **Model Question Paper**

#### **QP CODE:**

Reg No: \_\_\_\_\_

Name: \_\_\_\_\_

PAGES:4

## APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

#### SIXTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

#### **Course Code: AIT 322**

#### **Course Name: Concepts in Computer Graphics and Image Processing**

Max.Marks:100

**Duration: 3 Hours** 

# PART A

Answer All Questions. Each Question Carries 3 Marks

- 5. Justify the approach of using integer arithmetic in Bresenham's line drawing algorithm.
- 6. Consider a raster system with a resolution of 1024\*1024. What is the size of the raster needed to store 4 bits per pixel? How much storage is needed if 8 bits per pixel are to be stored?
- 7. Show that two successive reflections about either of the coordinate axes is equivalent to a single rotation about the coordinate origin.
- **8.** Determine a sequence of basic transformations that are equivalent to the x-direction shearing matrix.
- **9.** Find the window to viewport normalization transformation with window lower left corner at (1,1) and upper right corner at (2,6).
- **10.** Find the orthographic projection of a unit cube onto the x=0, y=0 and z=0 plane.
- 11. Define Sampling and Quantization of an image.
- **12.** Give any three applications of digital image processing.
- **13.** A captured image appears very dark because of wrong lens aperture setting. Describe an enhancement technique which is appropriate to enhance such an image.
- 14. Suggest an approach of thresholding that should be used in case of uniform illumination.

(10x3=30)

#### Part B

#### (Answer any one question from each module. Each question carries 14 Marks)

- **15.** (a) Write Midpoint circle drawing algorithm and use it to plot a circle with radius=20 and center is (50,30). (10)
  - (b) Draw the architecture of raster scan display systems and explain its working (4) principle.

#### OR

**16.** (a) Derive the initial decision parameter of Bresenham's line drawing algorithm (10) and use the algorithm to rasterize a line with endpoints (2,2) and (10,10).

	(b)	Explain the working principle of color CRT monitors with suitable illustrations.	(4)
17.	(a)	Compare boundary fill algorithm and flood fill algorithm.	(5)
	(b)	Reflect a triangle ABC about the line $3x-4y+8=0$ . The position vector of the coordinate ABC is given as A(4,1), B(5,2) and C(4,3). <b>OR</b>	(9)
18.	(a)	Explain the need of using vanishing points in projections.	(4)
	(b)	Explain Cohen-Sutherland line clipping algorithm. Use the algorithm to clip line P1(70, 20) and P2(100,10) against a window lower left hand corner (50,10) and upper right hand corner (80,40).	(10)
19.	(a)	Describe Sutherland Hodegman polygon clipping algorithm and what are its limitations.	(7)
	(b)	Explain how visible surfaces can be detected using depth buffer algorithm.	(7)
		OR	
20.	(a)	Describe Sutherland Hodegman polygon clipping algorithm and what are its limitations.	(7)
	(b)	Explain how visible surfaces can be detected using depth buffer algorithm.	(7)
21.	(a)	Explain the components of an image processing system with suitable diagram	(9)
	(b)	Define Resolution of an image. Explain the spatial and gray level resolution of an image with an example.	(5)
		OR	
22.	(a)	Define 4-adjacency, 8 adjacency and m-adjacency. Consider the image segment shown. 4 2 3 2 (q) 3 3 1 3	(7)

- 2 3 2 2
  - (p) 2 1 2 3

Let  $V=\{1,2\}$  and compute the length of the shortest 4- ,8- and m- path between p and q. If a particular path does not exist between these two points, explain why?

- (b) Using any one application, explain the steps involved in image processing. (7)
- 23. (a) A 5x5 image patch is shown below. Compute the value of the marked pixel if (4) it is smoothened by a 3x3 average filterand median filter.

f (m,n)=	0	1	2	3	2	
	5	6	7	8	4	
	4	3	2	1	2	
	8	7	6	5	3	
	1	5	3	7	6	
	$\mathcal{L}$					

(b) Define Image segmentation and describe in detail method of edge and region (10) based segmentation technique.

# OR

24.	(a)	Distin	iguish b	etween smoothing and sharpening filters in terms of	(10)
			5.	Functionality	
			6.	Types	
			7.	Applications	
			8.	Mask Coefficients	
	(b)	Descr	ibe how	an image is segmented using split and merge technique in	(8)

(b) Describe how an image is segmented using split and merge technique in association with the region adjacency graph.

# **Teaching Plan**

No	Contents	No of Lecture Hrs (36 hrs)
	Module – 1 (Basics of Computer Graphics and Algorithms) (8 hr	s)

	B TECH ARTIFICIAL INTELLIGENCE AND MACHI	
1.1	Basics of Computer Graphics and applications	1 hour
1.2	Refresh Cathode Ray Tubes	1 hour
1.3	Random Scan Displays and systems, Raster scan displays and systems	1 hour
1.4	DDA Line drawing Algorithm	1 hour
1.5	Bresenham's line drawing algorithm	1 hour
1.6	Midpoint Circle generation algorithm	1 hour
1.7	Bresenham's Circle generation algorithm	1 hour
1.8	Illustration of line drawing and circle drawing algorithms	1 hour
	Module - 2 (Filled Area Primitives and transformations) (8 hrs)	
2.1	Scan line polygon filling	1 hour
2.2	Boundary filling and flood filling	1 hour
2.3	Basic 2D transformations-Translation, Rotation and Scaling	1 hour
2.4	Reflection and Shearing	1 hour
2.5	Composite transformations	1 hour
2.6	Matrix representations and homogeneous coordinates	1 hour
2.7	Basic 3D transformation-Translation and scaling	1 hour
2.8	Basic 3D transformation-Rotation	1 hour
	Module - 3 (Clipping and Projections) (7 hrs)	
3.1	Window to viewport transformation	1 hour
3.2	Cohen Sutherland Line clipping algorithm	1 hour
3.3	Sutherland Hodgeman Polygon clipping algorithm	1 hour
3.4	Practice problems on Clipping algorithms	1 hour
3.5	Three-dimensional viewing pipeline, Projections-Parallel projections, Perspective projections	1 hour
3.6	Visible surface detection algorithms- Depth buffer algorithm	1 hour
3.7	Scan line visible surface detection algorithm	1 hour

	Module - 4 (Fundamentals of Digital Image Processing) (6 hrs)	
4.1	Introduction to Image processing-Image as a 2D data, Image representation-Gray scale, Binary and Colour images.	1 hour
4.2	Fundamental steps in image processing and applications	1 hour
4.3	Components of image processing system	1 hour
4.4	Coordinate conventions, Sampling and quantization, Spatial and Gray Level Resolution	1 hour
4.5	Basic relationship between pixels – neighbourhood, adjacency, connectivity	1 hour
4.6	Illustration of basic relationship between pixels- neighbourhood, adjacency, connectivity	1 hour
<b>Mo</b> 5.1	Iule - 5 (Image Enhancement in spatial domain and Image Segmentation         Basic gray level transformation functions- Log transformations, Power	
		1 hour
5.2	Basic gray level transformation functions- Log transformations, Power law transformation, Contrast stretching         Histogram equalization with illustration	1 hour 1 hour
5.2 5.3	law transformation, Contrast stretching	
	law transformation, Contrast stretching         Histogram equalization with illustration         Basics of spatial filtering, Smoothing spatial filter- Linear and	1 hour
5.3	law transformation, Contrast stretching         Histogram equalization with illustration         Basics of spatial filtering, Smoothing spatial filter- Linear and nonlinear filters	1 hour 1 hour
5.3 5.4	Iaw transformation, Contrast stretchingHistogram equalization with illustrationBasics of spatial filtering, Smoothing spatial filter- Linear and nonlinear filtersSharpening spatial filtering-Gradient filter mask, Laplacian Filter MaskFundamentals of Image Segmentation, Basics of Intensity thresholding,	1 hour 1 hour 1 hour

CST 332	FOUNDATIONS OF SECURITY IN	Category	L	Т	Р	Credit	Year Of Introduction
332	COMPUTING	PEC	2	1	0	3	2019

**Preamble:** The purpose of this course is to create awareness among learners about the fundamentals of security and number theory. This course covers Integer & Modular Arithmetic, Primes & Congruences, Discrete Logarithms & Elliptic Curve Arithmetic and an overview of computer security. The concepts covered in this course enable the learners in effective use of cryptographic algorithms and to identify the security threats in computing.

**Prerequisite:** A sound knowledge in Mathematics, Discrete Computational Structures, Operating Systems and Database Systems.

Course Outcomes: After the completion of the course, the student will be able to

CO1	Illustrate the operations and properties of algebraic structures, integer arithmetic and modular arithmetic. (Cognitive Knowledge Level: Understand)
CO2	Use the concepts of prime numbers and factorization for ensuring security in computing systems (Cognitive Knowledge Level: Apply)
CO3	Illustrate the concepts of Linear Congruence, Primitive Roots, Discrete Logarithms and Elliptic Curve Arithmetic (Cognitive Knowledge Level: Apply)
CO4	Summarize the threats and attacks related to computer and program security (Cognitive Knowledge Level: Understand)
CO5	Outline the key aspects of operating system and database security (Cognitive Knowledge Level: Understand)

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO1 0	PO1 1	PO1 2
CO1	9	0	0									9
CO2	0	0	9	0	D	JI	. K	A	A	Μ		0
CO3	0	0	0	0	1		$\bigcirc$		C/			0
CO4	0	0	9	N	Ŵ	0	S	0	Y			0
CO5	٢	0	0			0		0				۲

# Mapping of course outcomes with program outcomes

		Abstract POs defined by Nat	ional Bo	oard of Accreditation
PO#		Broad PO	PO#	Broad PO
PO1	Engine	eering Knowledge	PO7	Environment and Sustainability
PO2	Proble	m Analysis	PO8	Ethics
PO3	Design	n/Development of solutions	PO9	Individual and team work
PO4	Condu proble	ct investigations of complex ms	PO10	Communication
PO5	Moder	rn tool usage	PO11	Project Management and Finance
PO6	The E	ngineer and Society	PO12	Life long learning

# **Assessment Pattern**

Bloom's Category	Test 1 (%)	Test 2 (%)	End Semester Examination (%)
Remember	30	30	30
Understand	30	30	30
Apply	40	40	40
Analyse			

Evaluate		
Create		

#### **Mark Distribution**

Total Marks	<b>CIE Marks</b>	ESE Marks	ESE Duration
150	50	100	3

Continuous Ir	nternal Evaluation P	attern:
Attendance		: 10 marks
Continuous As	ssessment Test	: 25 marks

Continuous Assessment Assignment : 15 marks

# **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed modules and 1 question from the should answer all questions, a student should answer any 5.

# **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.

# **Syllabus**

#### Module-1 (Modular Arithmetic)

Integer arithmetic - Integer division, Divisibility, Greatest Common Divisor (GCD), Euclid's algorithm for GCD, Extended Euclid's algorithm, Linear Diophantine Equations. Modular

arithmetic - Operations, Properties. Algebraic structures - Groups, Rings, Fields, Finite fields, GF(p),  $GF(2^n)$ .

#### Module-2 (Prime Numbers and Factorization)

Prime numbers - Prime numbers and prime-power factorization, Fermat and Mersenne primes, Fermat's theorem, Applications, Euler's theorem, Euler's totient function, Applications. Primality testing – Deterministic algorithms and Probabilistic algorithms. Factorization - Fermat's factorization, Pollard p-1 method.

#### Module-3 (Linear Congruence, Primitive Roots and Elliptic Curve Arithmetic)

Linear congruence - Simultaneous linear congruence, Chinese Remainder Theorem (CRT). Congruence with a prime - Power modulus, Arithmetic modulo p, Pseudoprimes and Carmichael numbers, Solving congruence modulo prime powers. Primitive roots - Existence of primitive roots for primes, Discrete logarithms. Elliptic curve arithmetic – Prime curves, Binary curves, Addition of two points, Multiplication of a point by a constant.

#### Module-4 (Computer and Program Security)

Introduction to computer security – Threats, Vulnerabilities, Controls. Browser attack types, Web attacks targeting users, Email attack types. Introduction to program security - Non-malicious programming oversights, Malware.

#### Module-5 (Operating System and Database Security)

Operating system security – Security in operating system, Security in design of operating system. Database security – Security requirements of databases, Reliability and integrity, Database disclosure.

#### **Text Books**

- 5. Behrouz A Forouzan, Cryptography and Network Security, 3/e, Tata McGraw-Hill.
- 6. Charles P Pfleeger, Shari Lawrence Pfleeger, Jonathan Margulies, Security in Computing, 5/e, Prentice Hall.
- 7. G.A. Jones & J.M. Jones, Elementary Number Theory, Springer UTM, 2007

# References

a. William Stallings, Cryptography and Network Security Principles and Practices, 4/e, Pearson Ed.

## Sample Course Level Assessment Questions

## **Course Outcome 1 (CO1):**

- 6. Find the n- bit word that is represented by the polynomial  $x^2 + 1$  in GF(2<sup>5</sup>).
- 7. Solve the linear Diophantine equation 21x + 14y=35.

# **Course Outcome 2 (CO2):**

- 21. Prove that a Carmichael number cannot be the product of two distinct primes.
- 22. Use the Pollard p-1 method to find a factor of 57247159 with the bound B=8.

# **Course Outcome 3 (CO3):**

- 4. Find an integer that has a remainder of 3 when divided by 7 and 13, but is divisible by 12.
- 5. In the elliptic curve E(1,2) over the field GF(11), find the equation of the curve and all the points on the curve.

#### **Course Outcome 4 (CO4):**

- 5. List three controls that could be applied to detect or prevent off-by-one errors.
- 6. How does fake email messages act as spam?

#### Course Outcome 5 (CO5):

- 1. Discuss the importance of auditability and access control in database security.
- 2. Explain the various factors which can make data sensitive.

#### **Model Question Paper**

## **QP CODE:**

PAGES:

Reg No:\_\_\_\_\_ Name:\_\_\_\_\_

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

# SIXTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

#### Course Code: CST 332 Course Name : FOUNDATIONS OF SECURITY IN COMPUTING

#### Max Marks: 100

**Duration: 3 Hours** 

# PART A (Answer All Questions. Each question carries 3 marks)

- 1. List the four properties of divisibility with examples.
- 2. Find gcd (401,700) using Euclid's algorithm.
- 3. Use Fermat's Little theorem to show that 91 is not a prime.
- 4. If m is relatively prime to n, show that  $\Phi(mn) = \Phi(m) \Phi(n)$ .
- 5. Solve the congruence relation  $103x \equiv 57 \pmod{211}$ .
- 6. Find a solution for the congruence  $3x \equiv 5 \mod 7^3$
- 7. What are the problems created by an off-by-one error?
- 8. How does a clickjacking attack succeed?
- 9. Explain the significance of correctness and completeness in the design of operating systems.
- 10. How does the two-phase update technique help the database manager in handling failures? (10x3=30)

#### Part B

# (Answer any one question from each module. Each question carries 14 Marks)

5.	(a)	For the group $G = \langle Z_6^*, x \rangle$ , prove that it is an Abelian group. Also show the result of 5 x 1 and 1 ÷ 5.	(6)
	•	Find a particular and the general solution to the following linear Diophantine equations.	(8)
		i) $19 x+13y = 20$ ii) $40 x + 16 y = 88$	(0)
		AP AR OR L KALAM	
6.	(a)	Describe the properties of modular arithmetic and modulo operator.	(6)
	•	Using Extended Euclidean algorithm, find the multiplicative inverse of (i) 131 in $Z_{180}$ and (ii) 23 in $Z_{100}$ .	(0) (8)
7.		(a) State and prove Fermat's theorem.	(6)
	•	Explain Fermat's factorization method and use it to factor 809009.	(8)
		OR	
8.	(a)	Define Euler's totient function. Prove that, Ø(pq)=(p-1)(q-1) where p and q are prime numbers.	(7)
	•	Define Fermat primes. Show that any two distinct Fermat numbers are relatively prime.	(7)
9.	(a)	Using Chinese Remainder Theorem, solve the system of congruence, $x \equiv 2 \pmod{3}$ , $x \equiv 3 \pmod{5}$ , $x \equiv 2 \pmod{7}$ .	(7)
	•	Define Carmichael number and show that a Carmichael number must be the product of at least three distinct primes.	(7)
		OR	
10.	(a)	For the group $G = \langle Z_{19^*}, x \rangle$ , find the primitive roots in the group.	(6)
	•	Consider the elliptic curve $y^2 = x^3 + x + 1$ defined over Z <sub>23</sub> . If P = (3, 10) and Q = (9,7) are two points on the elliptic curve, find 2P and P + Q.	(8)
11.	(a)	Distinguish the terms vulnerability, threat and control.	(4)
	•	With the help of suitable examples, explain the security problems created by incomplete mediation and time-of-check to time-of use.	(10)
		OR	
12.	(a)	Differentiate between man-in-the-browser attack and page-in-the-middle	(4)

attack.

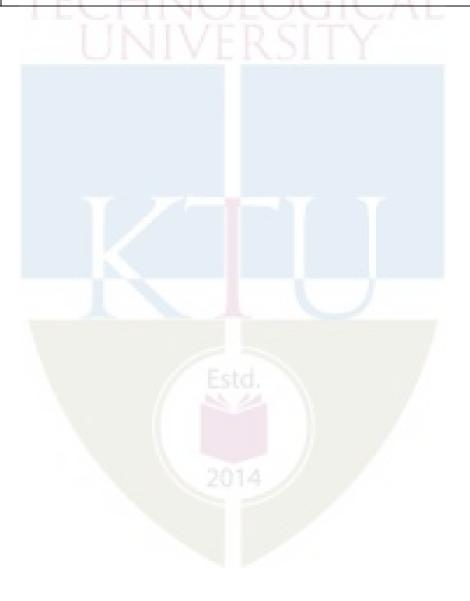
	(b)	Explain the four aspects of malicious code infection.	(10)
19.	(a)	List any six computer security related functions addressed by operating systems.	(6)
	(b)	How does a kernelized design support in enforcing security mechanisms?	(8)
		APJ ABDLOR KALAM	
20.	(a)	Explain any four security requirements of databases.	(4)
	(b)	How can database disclosure be prevented? With the help of suitable examples, explain any six types of disclosure.	(10)

# **Teaching Plan**

No	Contents	No. of Lecture Hrs
	Module-1 (Mod <mark>ul</mark> ar Arithmetic) (6 hrs)	
1.1	Integer arithmetic, Integer division, Divisibility, Greatest Common Divisor (GCD)	1
1.2	Euclid's algorithm for GCD, Extended Euclid's algorithm	1
1.3	Linear Diophantine Equations	1
1.4	Modular arithmetic operations, Properties of modular arithmetic	1
1.5	Groups, Rings and Fields	1
1.6	Finite fields – $GF(p)$ , $GF(2^n)$	1
	Module-2 (Prime Numbers and Factorization) (7 hrs)	·
2.1	Prime numbers and prime-power factorization	1
2.2	Fermat and Mersenne primes	1
2.3	Fermat's theorem, Applications – Exponentiation, Multiplicative inverse	1
2.4	Euler's theorem, Euler's totient function, Applications	1
2.5	Primality testing – Deterministic algorithms – Divisibility algorithm	1

2.6	Primality testing – Probabilistic algorithms-Fermat test, Square root test, Miller - Rabin test	1
2.7	Factorization - Fermat's factorization, Pollard p-1 method	1
Modu	lle-3 (Linear Congruence, Primitive Roots and Elliptic Curve Arithmetic	e) (7 hrs)
3.1	Linear congruence, Simultaneous linear congruence	1
3.2	Chinese Remainder Theorem (CRT)	1
3.3	Congruence with a Prime-Power Modulus, Arithmetic modulo p	1
3.4	Pseudo-primes and Carmichael numbers	1
3.5	Solving congruence modulo prime powers	1
3.6	Primitive roots, Existence of primitive roots for primes, Discrete logarithms	1
3.7	Elliptic curve arithmetic – Prime curves, Binary curves, Addition of two points, Multiplication of a point by a constant	1
Mo	Jule 4 (Computer and Program Security) (7 hrs) (Text heals): Chapters	1 2 4
11200	lule-4 (Computer and Program Security) (7 hrs) (Text book2: Chapters	1, 3, 4)
4.1	Threats, Vulnerabilities, Controls	1, 3, 4) 1
4.1	Threats, Vulnerabilities, Controls	1
4.1 4.2	Threats, Vulnerabilities, Controls       Browser attack types	1
<ul><li>4.1</li><li>4.2</li><li>4.3</li></ul>	Threats, Vulnerabilities, Controls         Browser attack types         Web attacks targeting users	1 1 1
<ul><li>4.1</li><li>4.2</li><li>4.3</li><li>4.4</li></ul>	Threats, Vulnerabilities, Controls         Browser attack types         Web attacks targeting users         Email attack types	1 1 1 1
<ul> <li>4.1</li> <li>4.2</li> <li>4.3</li> <li>4.4</li> <li>4.5</li> </ul>	Threats, Vulnerabilities, Controls         Browser attack types         Web attacks targeting users         Email attack types         Non-malicious programming oversights (Lecture 1)	1 1 1 1 1
<ul> <li>4.1</li> <li>4.2</li> <li>4.3</li> <li>4.4</li> <li>4.5</li> <li>4.6</li> <li>4.7</li> </ul>	Threats, Vulnerabilities, Controls         Browser attack types         Web attacks targeting users         Email attack types         Non-malicious programming oversights (Lecture 1)         Non-malicious programming oversights (Lecture 2)	1 1 1 1 1 1 1
<ul> <li>4.1</li> <li>4.2</li> <li>4.3</li> <li>4.4</li> <li>4.5</li> <li>4.6</li> <li>4.7</li> </ul>	Threats, Vulnerabilities, ControlsBrowser attack typesWeb attacks targeting usersEmail attack typesNon-malicious programming oversights (Lecture 1)Non-malicious programming oversights (Lecture 2)Malware – Four aspects of infection	1 1 1 1 1 1 1
4.1 4.2 4.3 4.4 4.5 4.6 4.7 Module	Threats, Vulnerabilities, Controls         Browser attack types         Web attacks targeting users         Email attack types         Non-malicious programming oversights (Lecture 1)         Non-malicious programming oversights (Lecture 2)         Malware – Four aspects of infection         e-5 (Operating System and Database Security) (8 hrs)(Text book2: Chapter 1)	1 1 1 1 1 1 1 ters 5, 7

5.4	Security in design of operating system (Lecture 2)	1
5.5	Security requirements of databases	1
5.6	Reliability & integrity	1
5.7	Database disclosure (Lecture 1)	1
5.8	Database disclosure (Lecture 2)	1



AIT 342	Concepts in Formal Languages and	Category	L	Т	Р	Credit	YEAR OF INTRODUCTION
342	Automata Theory	PEC	2	1	0	3	2020

**Preamble**: This is a core course in theoretical computer science. It covers automata and grammar representations for languages in Chomsky Hierarchy. For regular languages, it also covers representations using regular expression and Myhill-Nerode Relation. The topics covered in this course have applications in various domains including compiler design, decidability and complexity theory, software testing, formal modelling and verification of hardware and software.

# Prerequisite: Basic knowledge about the following topic is assumed: sets, relations - equivalence relations, functions, proof by Principle of Mathematical Induction.

**Course Outcomes**: After the completion of the course the student will be able to

CO1	Classify a given formal language into Regular, Context-Free, Context Sensitive, Recursive or Recursively Enumerable. [Cognitive knowledge level: Understand]
CO2	Explain a formal representation of a given regular language as a finite state automaton, regular grammar, regular expression and Myhill-Nerode relation. [Cognitive knowledge level:Understand]
CO3	Design a Pushdown Automaton and a Context-Free Grammar for a given context- free language. [Cognitive knowledge level: Apply]
CO4	Design Turing machines as language acceptors or transducers. [Cognitive knowledge level: Apply]
CO5	Explain the notion of decidability. [Cognitive knowledge level: Understand]

#### Mapping of course outcomes with program outcomes

	PO1	PO2	PO3	PO4	PO5	PO6	<b>PO7</b>	PO8	PO9	PO1 0	PO11	PO1 2
CO1												
CO2	$\bigcirc$	$\bigcirc$	$\bigcirc$	$\bigcirc$								$\bigcirc$
CO3	$\bigcirc$	$\bigcirc$	$\bigcirc$	$\bigcirc$								

CO4	$\oslash$	$\oslash$	$\oslash$	$\bigcirc$							
CO5	$\oslash$	٢	٢	0							
		Al	1	٩B	D	. K	A	A	M	<u>.</u>	

	Abstract POs defined by Na	ational B	Board of Accreditation
PO#	Broad PO	PO#	Broad PO
PO1	Engineering Knowledge	PO7	Environment and Sustainability
PO2	Problem Analysis	PO8	Ethics
PO3	Design/Development of solutions	PO9	Individual and team work
PO4	Conduct investigations of complex problems	PO10	Communication
PO5	Modern tool usage	PO11	Project Management and Finance
PO6	The Engineer and Society	PO12	Life long learning

# **Assessment Pattern**

Bloom's Category	Continuous A	End Semester	
	Test 1 (Marks)	Test 2 (Marks)	Examination Marks
Remember	30	30	30
Understand	30	30	30
Apply	40	40	40
Analyze			
Evaluate			
Create			

# **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
-------------	-----------	-----------	--------------

150	50	100	3 hours
<b>Continuous Internal Eva</b> Attendance	luation Pattern:	: 10 marks	
Continuous Assessment -	Гest	: 25 marks	
Continuous Assessment - 2	Assignment	: 15 marks	

#### **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. The first series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing the remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed modules and 1 question from the should answer all questions, a student should answer any 5.

#### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.

#### **Syllabus**

#### **CST 301 Formal Languages and Automata Theory**

Module - 1 (Introduction to Formal Language Theory and Regular Languages)

Regular Languages - Deterministic Finite State Automata (DFA) (Proof of correctness of construction not required), Nondeterministic Finite State Automata (NFA), Equivalence of DFA and NFA, Regular Grammar (RG), Equivalence of RGs and DFA.

# Module - 2 (More on Regular Languages)

Regular Expression (RE), Equivalence of REs and DFA, Necessary conditions for regular languages, Closure Properties of Regular Languages, DFA state minimization (No proof required).

# Module - 3 (Myhill-Nerode Relations and Context Free Grammars)

Myhill-Nerode Relations (MNR)- MNR for regular languages, Myhill-Nerode Theorem (MNT) (No proof required), Applications of MNT.

Context Free Grammar (CFG)- CFG representation of Context Free Languages (proof of correctness is required), derivation trees and ambiguity, Normal forms for CFGs.

# Module - 4 (More on Context-Free Languages)

Nondeterministic Pushdown Automata (PDA), Equivalence of PDAs and CFGs (Proof not required), Pumping Lemma for Context-Free Languages (Proof not required), Closure Properties of Context Free Languages.

# Module - 5 (Context Sensitive Languages, Turing Machines)

Context Sensitive Languages - Context Sensitive Grammar (CSG).

Turing Machines - Standard Turing Machine, Universal Turing Machine, Halting Problem, Recursive and Recursively Enumerable Languages.

Chomsky classification of formal languages.

# **Text Book**

6. Dexter C. Kozen, Automata and Computability, Springer (1999)

# **Reference Materials**

- 6. John E Hopcroft, Rajeev Motwani and Jeffrey D Ullman, Introduction to Automata Theory, Languages, and Computation, 3/e, Pearson Education, 2007
- 7. Michael Sipser, Introduction To Theory of Computation, Cengage Publishers, 2013.

# Sample Course Level AssessmentQuestions

Course Outcome 1 (CO1): Identify the class of the following languages in Chomsky Hierarchy:

- 6.  $L_1 = \{a^p | pis \ a \ prime \ number\}$
- 7  $L_2 = \{x\{0,1\}^* | xis the binary representation of a decimal number which is a multiple of 5\}$

- 8.  $L_3 = \{a^n b^n c^n | n \ge 0\}$
- 9  $L_4 = \{a^m b^n c^{m+n} | m > 0, n \ge 0\}$
- 10.  $L_5 = \{M \# x | M halts on x\}$ . Here, M is a binary encoding of a Turing Machine and x is a binary input to the Turing Machine.

## **Course Outcome 2 (CO2):**

- (i) Design a DFA for the language  $L = \{axb | x \in \{a,b\}^*\}$
- (ii) Write a Regular Expression for the

language:  $L = \{x \in \{a,b\}^* | third \ last \ symbol \ in \ x \ is \ b\}$ 

(iii) Write a Regular Grammar for the

language:  $L = \{x \in \{0,1\}^* | there are no consecutive zeros inx\}$ 

(iv) Show the equivalence classes of the canonical Myhill-Nerode relation induced by the language:  $L = \{x \in \{a,b\}^* | x \text{ contains even number of } a \text{ 's and odd number of } b \text{ 's} \}$ .

#### **Course Outcome 3 (CO3):**

- (i) Design a PDA for the language  $L = \{ww^R | w \in \{a,b\}^*\}$ . Here, the notation  $w^R$  represents the reverse of the string w.
- (ii) Write a Context-Free Grammar for the language  $L = \{a^n b^{2n} | n \ge 0\}$ .

# **Course Outcome 4 (CO4):**

- (i) Design a Turing Machine for the language  $L = \{a^n b^n c^n | n \ge 0\}$
- (ii) Design a Turing Machine to compute the square of a natural number. Assume that the input is provided in unary representation.
- **Course Outcome 5 (CO5):** Argue that it is undecidable to check whether a Turing Machine Menters a given state during the computation of a given input x.

#### **Model Question paper**

**QP CODE:** 

PAGES:3

Name :

Reg No:\_\_\_\_\_

APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

# FIFTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR Course Code: AIT 342

**Course Name: Concepts in Formal Languages and Automata Theory** 

# Max.Marks:100

**Duration: 3 Hours** 

## PART A

# Answer all Questions. Each question carries 3 Marks

- 1. Design a DFA for the language  $L = \{x \in \{a,b\}^* | aba \text{ is not a substring in } x\}$ .
- 2. Write a Regular Grammar for the language:  $L = \{axb | x \in \{a,b\}^*\}$
- 3. Write a Regular Expression for the language:  $L = \{x \in \{0,1\}^* | there are no consecutive 1's in x\}$
- 4. Prove that the language  $L_1 = \{a^{n!} | n \in N\}$  is not regular.
- 5. List out the applications of Myhill-Nerode Theorem.
- 6. Write a Context-Free Grammar for the language:  $L = \{x \in \{a,b\}^* | \#_a(x) = \#_b(x)\}$ . Here, the notation  $\#_1(w)$  represents the number of occurrences of the symbol 1 in the string w.
- 7. Design a PDA for the language of odd length binary palindromes (no explanation is required, just list the transitions in the PDA).
- 8. Prove that Context Free Languages are closed under set union.
- 9. Write a Context Sensitive Grammar for the language  $L = \{a^n b^n c^n | n \ge 0\}$  (no explanation is required, just write the set of productions in the grammar).
- 10. Differentiate between Recursive and Recursively Enumerable Languages.

(10x3=30)

# Part B (Answer any one question from each module. Each question carries 14 Marks)

11. (a) Draw the state-transition diagram showing an NFA N for the following language L. Obtain the DFA<sup>D</sup> equivalent to N by applying the subset (7) construction algorithm.

$$L = \{x \in \{a,b\}^* | the second last symbol in x is b\}$$

(b) Draw the state-transition diagram showing a DFA for recognizing the following language:

 $L = \{x \in \{0,1\}^* | x \text{ is a binary representation of a natural} \}$ number which is a<sub>multiple of 5</sub>}

#### OR

- 12. (a) Write a Regular grammar G for the following language L defined as:  $L = \{x \in \{a,b\}^* | x does not contain consecutive b's\}$ (7)
  - (b) Obtain the DFA  $A_G$  over the alphabet set  $\Sigma = \{a, b\}$ , equivalent to the regular (7) grammar  $^{G}$  with start symbol  $^{S}$  and productions:  $^{S} \rightarrow aA$  and  $^{A} \rightarrow aA|bA|b$
- 13. (a) Using Kleen's construction, obtain the regular expression for the language represented by the following NFA

(8)

(b) Using pumping lemma for re  $L = \{a^n b^n | n \ge 0\}$  is not regul

OR

14. (a) the

a, b

Obtain (8)

(7)

(7)

minimum-state DFA from the following DFA.

	(b)	Using ultimate periodicity for regular languages, prove that the language $L = \{a^{n^2}   n \ge 0\}$ is not regular.	(6)
15.	(a)	Show the equivalence classes of the canonical Myhill-Nerode relation for the language of binary strings with odd number of $1$ 's and even number of $0$ s.	(7)
	(b)	With an example, explain ambiguity in Context Free Grammar	(7)
		OR	
16.	(a)	Convert the Context-Free Grammar with productions: $\{S \rightarrow aSb   \epsilon\}$ into Greibach Normal form.	(8)
	(b)	Convert the Context-Free Grammar with productions: $\{S \rightarrow aSa   bSb   SS   \epsilon\}$ into Chomsky Normal form.	(6)
17.	(a)	Design a PDA for the language $L = \{a^m b^n c^{m+n}   n \ge 0, m \ge 0\}$ . Also illustrate the computation of the PDA on a string in the language	(7)
	(b)	With an example illustrate how a multi-state PDA can be transformed into an equivalent single-state PDA.	(7)
		OR	
18.	(a)	Using pumping lemma for context-free languages, prove that the language: $L = \{ww   w \in \{a,b\}^*\}$ is not a context-free language.	(6)
	(b)	With an example illustrate how a CFG can be converted to a single-state PDA	(8)
19.	(a)	Design a Turing machine to obtain the sum of two natural numbers $a_{and} b$ , both represented in unary on the alphabet set $\{1\}$ . Assume that initially the	(7)

tape contains  $\vdash 1^a 01^b \models {}^{\omega}$ . The Turing Machine should halt with  $\vdash 1^{a+b} \models {}^{\omega}$  as the tape content. Also, illustrate the computation of your Turing Machine on the input  $a = 3_{\text{and}} b = 2_{\text{constrained}}$ .

(b) With an example illustrate how a CFG can be converted to a single-state (7) PDA.

# OR

- 20. (a) Design a Turing machine to obtain the sum of two natural numbers and b, (7) both represented in unary on the alphabet set {1}. Assume that initially the tape contains ⊢ 1<sup>a</sup>01<sup>b</sup> b<sup>ω</sup>. The Turing Machine should halt with ⊢ 1<sup>a+b</sup> b<sup>ω</sup> as the tape content. Also, illustrate the computation of your Turing Machine on the input a = 3<sub>and</sub> b = 2.
  - (b) Write a context sensitive grammar for the language  $L = \{a^n b^n c^n | n \ge 0\}$ . Also (7) illustrate how the the string  $a^2 b^2 c^2$  can be derived from the start symbol of the proposed grammar.

# **Teaching Plan**

Sl. No	Торіс				
Mod	ule - 1 (Introduction to Formal Language Theory and Regular Languages)	8 Hours			
1.1	Deterministic Finite State Automata (DFA) – Example DFA (Proof of correctness of construction not required)	1 Hour			
1.2	Formal definition of DFA, Language accepted by the class of DFA	1 Hour			
1.3	Nondeterministic Finite State Automata (NFA) – Example NFA	1 Hour			
1.4	Formal definition of NFA, NFA with $\epsilon$ transitions - examples, formal definition	1 Hour			

1.5	Equivalence of DFA and NFA with and without $\epsilon$ transitions - Subset construction	1 Hour
1.6	Regular Grammar (RG) – Example RGs, derivation of sentences	1 Hour
1.7	Formal definition of RG, Language represented by a RG	1 Hour
1.8	Equivalence of RG and DFA	1 Hour
	Module - 2 (More on Regular Languages)	7 Hours
2.1	Regular Expression (RE) - Example REs and formal definition	1 Hour
2.2	Conversion of RE to NFA with $\epsilon$ transition	1 Hour
2.3	Conversion of NFA with $\epsilon$ transition to RE (Kleen's construction)	1 Hour
2.4	Pumping Lemma for regular languages	1 Hour
2.5	Ultimate periodicity	1 Hour
2.6	Closure Properties of Regular Languages (proof not required)	1 Hour
2.7	DFA state minimization - Quotient construction	1 Hour
	Module - 3 (Myhill-Nerode Relations and Context Free Grammars)	7 Hours
3.1	Myhill-Nerode Relations (MNR) - Example, Properties of MyhillNerode Relation	1 Hour
3.2	Conversion of DFA to MNR, Conversion of MNR to DFA (Proof of correctness not required)	1 Hour
3.3	Myhill-Nerode Theorem (MNT), Applications of MNT	1 Hour
3.4	Context Free Grammar (CFG) - Example CFGs and formal definition	1 Hour
3.5	Derivation Trees and ambiguity	1 Hour
3.6	Chomsky Normal Form	1 Hour
3.7	Greibach Normal Form	1 Hour
	Module - 4 (More on Context-Free Languages)	7 Hours
4.1	Nondeterministic Pushdown Automata (PDA) – Example PDAs, formal definition	1 Hour

4.2	Acceptance criteria - equivalence	1 Hour					
4.3	Conversion of CFG to PDA (No proof required)						
4.4	Conversion of PDA to CGF - Part I (No proof required)						
4.5	Conversion of PDA to CGF - Part II (No proof required)	1 Hour					
4.6	4.6 Pumping Lemma for context-free languages (No proof required)						
4.7	4.7 Closure Properties of Context Free Languages						
	Module - 5 (Context Sensitive Languages, Turing Machines )	7 Hours					
5.1	Context Sensitive Grammar (CSG) - Examples, formal definition	1 Hour					
5.2	Turing Machine (TM) - TM as language acceptors - examples, formal definition	1 Hour					
5.3	TM as transducers - examples	1 Hour					
5.4	Universal Turing Machine	1 Hour					
5.5	Halting Problem of TM - proof of its undecidability       1 Hour						
5.6	Recursive and Recursively Enumerable Languages	Recursive and Recursively Enumerable Languages 1 Hour					
5.7	Chomsky classification of formal languages	1 Hour					



AIT 352	Artificial Neural Networks Techniques	Category	L	Т	Р	Credit	Year of Introduction
	A DI A I	PEC	2	1	0	3	2019

# **Preamble**:

This course enables the learners to understand the fundamental concepts regarding Artificial Neural networks. The course covers basic analogy between ANN and human brain, the basic learning laws, fundamental ANN algorithms, Back Propagation Feed Forward Network, Self Organising Maps, RBF net, BAM and ART networks. This course enables the students to apply techniques and methods to solve real-world problems involving the application of ANN.

# Prerequisite: Nil.

# **Course Outcomes**

CO1	Summarize the basic concepts and the learning rules of ANN. ( <b>Cognitive Knowledge</b> Level: Understand)				
CO2	Utilize the fundamental learning algorithms namely, Mc-Culloch Pitts, Hebb Perceptron and Adaline to solve real world problems.(Cognitive Knowledge Level: Apply)				
CO3	Implement Back propagation learning algorithm, Generic Radial Basis Function network. (Cognitive Knowledge Level: Apply)				
CO4	Demonstrate Self Organizing Maps and Adaptive Resonance Theory.(Cognitive Knowledge Level: Understand)				
CO5	Implement training algorithms for pattern association. (Cognitive Knowledge Level: Apply)				

# Mapping of course outcomes with program outcomes

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1												Ø

CO2	$\bigcirc$	$\oslash$	Ø	$\oslash$	$\oslash$						$\bigcirc$
CO3		0	0	0	0						
CO4	Ø	0	1	A.E	0	Ų.	. K	A	A	Ŵ	$\bigotimes$
CO5	Ø	0	$\bigcirc$	0	0	DĽ.	0	10	Ţ	L	$\oslash$
		1	U	N	IV	EF	tS1		Y		

		Abstract POs defined by	Board of Accreditation	
PO#		Broad PO	PO#	Broad PO
PO1	Eng	ineering Knowledge	PO7	Environment and Sustainability
PO2	Prot	olem Analysis	PO8	Ethics
PO3	Dest	ign/Development of solutions	PO9	Individual and team work
PO4		duct investigations of complex plems	PO10	Communication
PO5	Mod	lern tool usage	PO11	Project Management and Finance
PO6	The	Engineer and Society	PO12	Life long learning

# **Assessment Pattern**

Bloom's Category	Continuous	s Assessment Tests	End Semester Examination Marks (%)
	Test 1 (%)	Test 2 (%)	
Remember	30	30	30
Understand	40	40	40
Apply	30	30	30
Analyze			
Evaluate			
Create			

15 marks

# **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3 hours
Continuous Internal Eva	duation Pattern:	LUGIC	AL
Attendance			10 marks

Continuous Assessment Assignment

# **Internal Examination Pattern**

Two internal examinations of two hours duration has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed modules, a student should answer any 5.

# **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 full questions from each module of which student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.

# **Syllabus**

# Module – 1 (Basics of Artificial Neural Network and Learning Methods)

Characteristics of the human brain, Neurons, Introduction to Artificial Neural Networks, Terminology, Models of ANN, Topology, Network Architectures, Knowledge Representation, Learning Process, Learning Tasks. Categories of learning - Hebbian learning, Perceptron Learning Rule, Delta Learning Rule, Generalized Delta Learning Rule, Competitive learning, Error-correction learning, Reinforcement learning, Stability and Convergence.

# Module – 2(Basic ANN Models)

McCulloch-Pitts Neuron, Architecture, Algorithm and Applications. Biases and Thresholds, Linear Separability. Hebb Net - Algorithm, Applications. Perceptron - Architecture, Algorithm, Applications. Perceptron Learning Rule Convergence Theorem. Adaline - Architecture, Algorithm, Applications.

# Module - 3 (Multilayer Perceptrons)

Multi-Layered network architecture, Back propagation Algorithm, Applications, XOR problem, Replacing and Modifying Back propagation Algorithms Using Heuristics.

Cover's Theorem on the Separability of patterns, The Interpolation Problem, Radial Basis Function Networks, Comparison of MLP and RBF Networks( Theory only).

# Module – 4 (SOMs and ART Networks)

Self-organizing maps - Building, Training, Evaluating, Interpreting and Visualizing a Self-organizing Map. Applications of Self Organizing Maps.

Adaptive Resonance Theory -Stability Plasticity Dilemma, ART-1-Architeture, Algorithm, Applications. ART-2 – Architeture, Algorithm, Applications.

# Module – 5 (Training Algorithms for Pattern Association)

Introduction, Hetero associative neural network- Architecture, Applications. Auto Associative Net -Architecture, Applications. Iterative Auto Associative Net – Architecture, Applications. Discrete Hopfield Network. Bidirectional Auto-associative Memory – Architecture, Applications.

# **Text Books**

- 1. Simon Haykin, "Neural Networks, A comprehensive Foundation" (2<sup>nd</sup> edition), Pearson Education (Module 4)
- 2. Laurene Faucett, "Fundamentals of Artificial Neural Networks, architecture algorithm and applications" (Modules 2,3,5)
- 3. Yegnanarayana, "Artificial Neural Networks", Phi Learning (Module -1)

# **Reference Books**

- 1. Christopher M Bishop," Neural networks for Pattern Recognition
- 2. Mohammad H Hassoun, "Fundamentals of Artificial Neural Networks"

# **Course Level Assessment Questions**

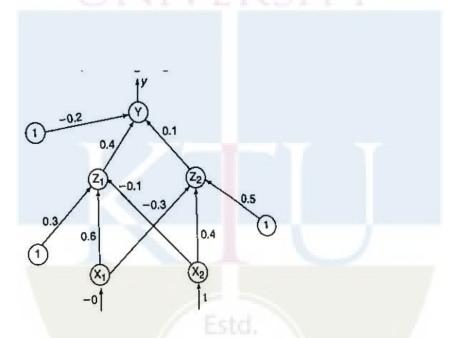
# **Course Outcome1 (CO1):**

- 1. What are the different types of competitive learning?
- **2.** Demonstrate the significance of different Activation functions.
- **3.** Explain the terms cell body, axon, synapse, dendrite and neuron with reference to abiological neural network.
- **4.** Illustrate examples of pattern recognition tasks to demonstrate the superiority of the biological neural network over a conventional computer system. (Assignment Question)

# Course Outcome 2 (CO2):

- **1.** How is training adopted in Adaline network and state the characteristics of weighted interconnections in Adaline .
- 2. How is the linear separability concept implemented using Perceptron Network training?
- **3.** Implement NAND logical function using Perceptron Network in Python language(Assignment Question)

Course Outcome 3(CO3):



- 1. Find the new weights of Back propagation net shown in the figure for the input pattern
  - (0,-1) and the target output 1, Use 0.25 as learning rate.
- 2. Why is gradient descent method adopted to minimize error? Explain in relation to Back propagation of error phase of BPNN?
- 3. Implement RBF network using Python language. (Assignment Question)

# Course Outcome 4(CO4): .

1. Design an ART1 used to cluster four vectors with low vigilance. The values and description of the parameters are given in the table. Cluster the vectors, (1,1,0,0), (0,0,0,1), (1,0,0,0), (0,0,1,1) in at most three clusters.

n=4	Number of components in the input vector
m=3	It was an excellent game.
P=0.4	Vigilence parameter
L=2	Parameter used in update of bottom-up weights
b <sub>ij</sub> (0)=1/n+1	Initial bottom-up weights

	]	]	]
--	---	---	---

- 2. Use NeuPy library of Python to implement Adaptive Resonance Theory (ART1) Network for binary data clustering.
- 3. Implement Self Organizing Map in Python to demonstrate how does the grid automatically arrange, using colour patterns and evaluate the effect of Learning Rate and Radius. (Assignment Question)

# **Course Outcome 5(CO5):**

- 1. Compare and contrast auto associative and hetero associative networks with examples.
- 2. Implement Bidirectional Associative Memory using Python without using specific libraries. (Assignment Question)

Model Question Paper

	Model Question 1 aper	
QP CODE:		
Reg No:		
Name:		PAGES:4

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

# SIXTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

Course Code: AIT 352

# **Course Name: Artificial Neural Networks Techniques**

Max. Marks: 100

**Duration: 3 Hours** 

# PART A

# Answer All Questions. Each Question Carries 3 Marks

1. What are the main differences among the three models of artificial neurons, namely, 3 McCulloch-Pitts, Perceptron and Adaline?

2.	Compare the stability and convergence of ANN	3
3.	Design a Mc-Culloch Pitts neural network to implement AND function.	3
4.	Define Perceptron Learning Rule Convergence theorem.	3
5.	What is the significance of momentum factor in backpropagation learning?	3
6.	Compare RBF network and Multilayer Perceptron network.	3
7.	Illustrate the feature mapping models.	3
8.	What is the significance of 'resonance' in ART network?	3
9.	Explain the hebb rule for pattern association	3
10.	Interpret cross talk and perfect recall using suitable examples	3

# Part B

# (Answer any one question from each module. Each question carries 14 Marks)

11.	(a)	Describe any four attractive features of the biological neural network that make it superior to the most sophisticated Artificial Intelligence computer system for pattern recognition tasks.	(8)
	(b)	Compare LMS, Perceptron and Delta learning laws.	(6)
		OR 014	
12.	(a)	Compare the performance of a computer and that of a biological neural network in terms of speed of processing, size and complexity, storage, fault tolerance and control mechanism.	(8)
	(b)	What is reinforcement learning? In what way it is different from supervised learning?	(6)
13.	(a)	Explain Hebb net algorithm and implement logical AND function using bipolar inputs.	(4)

	(b)	<ul> <li>b) Use Adaline nerwork to train AND NOT function with bipolar inputs and targets. Perform one epoch of training.</li> </ul>							
		APLARDI <sup>R</sup> KALAM							
14.	(a)	Using the Perceptron Learning rule find the weights required to perform the following classifications. Vectors $(1,1,1,1)$ and $(-1,1,-1,-1)$ are members of the class and hence target value 1; vectors $(1,1,1,-1)$ and $(1, -1, -1, 1)$ are not the members of the class and hence target value -1. Use learning rate of 1 and starting value of weights as 0, test the response of the net.	(10)						
	(b)	XOR function is non-linearly separable by a single decision boundary line. Justify.	(4)						
15.	(a)	Analyse Cover's theorem based on XOR problem.	(10)						
	(b)	Explain the learning factors of of Back propagation network algorithm	(4)						
16.	(a)	Relate Hidden layer and Output layer error terms with back propagation of error term phase in Back Propagation Network algorithm.	(10)						
	(b)	Explain the architecture and algorithm of RBF network .	(8)						
17.	(a)	Explain the statistical properties exhibited by SOM after convergence.	(10)						
	(b)	Interpret stability-plasticity dilemma in relation with ART network.	(4)						
		200R							
18.	(a)	Show the architecture of Kohonen's Self Organising Map and demonstrate the competitive process in Kohonen's self organising Map.	(8)						
	(b)	Explain the basic architecture of ART-2 and its algorithm.	(6)						
19.	(a)	Describe the architecture and algorithm of Discrete Bidirectional Associative Memory	(5)						

(b) Use the Hebb rule to store the vetors (1,1,1,1) and (1,1,-1,-1) in an auto i.(9) associative neural net.

- a. Fnd the weight matrix( Do not set the diagonal terms to zero)
- b. Test the net, using the following vectors as input
- i. (1,1,1,1)
- ii. (1,1,-1,-1)

iii. (1,1,1,0)

Repeat parts a and b with diagonal weight matrix set to zero. Identify the differences in the response.

#### OR

- 20. (a) Design a BAM net to associate the letters"A" and "C" given in bipolar 5X3 (10) vectors to the bipolar codes (-1,1) and (1,1) respectively.
  - (b) Compare Iterative Autoassocative with Discrete Hopfield Net. (4)

# **TEACHING PLAN**

No	Contents	No of Lecture Hrs: 35						
Mo	odule -1 : Basics of Artificial Neural Network and Learning methods	(7 hours)						
1.1	Introduction to Neural Network, The human brain - Characteristics of Neural Network.	1						
1.2	Artificial Neural Network - Terminology, Models of a neuron, Topology	1						
1.3	Network architectures, Knowledge representation.	1						
1.4	1.4 Learning Process, Learning tasks.Categories of learning- Hebbian learning, Competitive learning.							
1.5	Error-correction learning.	1						
1.6	Reinforcement learning.	1						
1.7	Stability and Convergence.	1						
	Module - 2 : Basic ANN Models( 7 hours)							

1								
2.1	McCulloch-Pitts Neuron - Architecture, Algorithm and Applications.	1						
2.2	Biases and thresholds, Linear separability.	1						
2.3	Hebb net - Algorithm, Applications	1						
2.4	Perceptron - Architecture, Algorithm	1						
2.5	Perceptron -Applications, Perceptron learning rule convergence theorem.	1						
2.6	Perceptron learning rule convergence theorem. Adaline - Architecture, Algorithm	1						
2.7	Adaline - Applications	1						
	Module 3 : Multilayer Perceptrons (7 hours)							
3.1	Multilayered Feed Forward Network Architecture,	1						
3.2	Back propagation algorithm, Activation functions, Rate of learning, Stopping criteria							
3.3	Applications, XOR problem, Heuristics for making the Back propagation algorithm perform better.							
3.4	Cover's Theorem on the separability of patterns.							
3.5	Cover's Theorem on the separability of patterns, XOR problem.	1						
3.6	The interpolation problem, Radial Basis Function networks.	1						
3.7	The interpolation problem, Radial Basis function networks, Comparison of RBF network and Multi-Layer perceptrons.       1							
	Module 4 : SOMs and ART networks (7 hours)							
4.1	Two basic feature mapping methods.	1						
4.2	Self Organizing Map, Competitive process, Cooperative process, Adaptive process.							
4.3	Properties of the feature map.	1						
4.4	Stability Plasticity Dilemma, ART-1-Architeture.	1						

4.5	ART-1 - Algorithm, Applications.	1					
4.6	ART-2 - Architeture-Algorithm	1					
4.7	ART-2 - Applications.	1					
	Module 5 : Training Algorithms for pattern Association (7 hour	s)					
5.1	Introduction, Hebb rule for pattern association, Delta rule for pattern association	1					
5.2	Hetero Associative Neural Network-Architecture, Applications, 1						
5.3	Auto-associative Net - Architecture, Algorithm, Applications, Storage capacity.	1					
5.4	Iterative Auto Associative Net - Architecture, Applications	1					
5.5	Discrete Hopfield network - Architecture, Algorithm, Applications.	1					
5.6	Bidirectional Auto-associative Memory-Architecture, Algorithm.	1					
5.7	Bidirectional Auto-associative Memory – Applications.	1					



AIT 362	PROGRAMMING IN R	CATEGORY	L	Т	P	CREDIT	YEAR OF INTRODUCTION
		PEC	2	1	0	3	2019

# **Preamble:**

The objective of this course is to enable the learner to make use of R Programming language to perform analysis and extraction of information from data irrespective of the quantity. It encompasses the R programming environment, syntax, data representations, data processing, statistical analysis and visualization. This course facilitates the learner to develop modular software solutions to perform statistical analysis and data extraction.

**Prerequisite:** Fundamental concepts in programming in C and Probability and Statistical Modeling

**Course Outcomes:** After the completion of the course the student will be able to:

CO 1	Illustrate uses of conditional and iterative statements in R programs. (Cognitive Knowledge level: Apply)
CO 2	Write, test and debug R programs (Cognitive Knowledge level: Apply)
CO 3	Illustrate the use of Probability distributions and basic statistical functions. (Cognitive Knowledge level: Apply)
CO 4	Visualize different types of data (Cognitive Knowledge level: Apply)
CO 5	Comprehend regression modeling using R (Cognitive Knowledge level: Understand)

# Mapping of course outcomes with program outcomes

	<b>PO1</b>	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1		$\bigcirc$										
CO2	$\bigotimes$				$\bigcirc$	20	114					$\bigotimes$
CO3												
CO4	$\bigcirc$	$\bigotimes$	$\bigcirc$	$\bigcirc$								
CO5	$\bigcirc$	$\bigcirc$			$\bigcirc$							

	Abstract POs defined by National Board of Accreditation										
PO#	Broad PO	PO#	Broad PO								
PO1	Engineering Knowledge	<b>PO7</b>	Environment and Sustainability								
PO2	Problem Analysis	PO8	Ethics								
PO3	Design/Development of solutions	PO9	Individual and team work								
PO4	Conduct investigations of complex problems	PO10	Communication								
PO5	Modern tool usage	PO11	Project Management and Finance								
PO6	The Engineer and Society	PO12	Life long learning								

# **Assessment Pattern**

	Continuous Assess	nent Tests	End	
Bloom's Category	Test1 (percentage)	Test2 (percentage)	Semester Examination Marks	
Remember	20	20	20	
Understand	40	40	40	
Apply	40	40	40	
Analyze				
Evaluate	Estd.			
Create				

# Mark distribution

Total	CIE	ESE	ESE
Marks	Marks	Marks	Duration
150	50	100	3 hours

# **Continuous Internal Evaluation Pattern:**

Attendance: 10 marks

Continuous Assessment Tests : 25 marks

Continuous Assessment Assignment: 15 marks

#### **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks

First Internal Examination shall be preferably conducted after completing the first half of the syllabus and the Second Internal Examination shall be preferably conducted after completing the remaining part of the syllabus.

There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly covered module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly covered module), each with 7 marks. Out of the 7 questions in Part B, a student should answer any 5.

#### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have a maximum of 2 subdivisions and carries 14 marks.

# Syllabus

#### Module -1 (Introduction to R)

The R Environment - Command Line Interface and Batch processing, R Packages, Variables, Data Types, Vectors- vector operations and factor vectors, List- operations, Data Frames, Matrices and arrays, Control Statements- Branching and looping - For loops, While loops, Controlling loops. Functions- Function as arguments, Named arguments

#### Module -2(Reading and writing data)

Importing data from Text files and other software, Exporting data, importing data from databases- Database Connection packages, Missing Data - NA, NULL

Combining data sets, Transformations, Binning Data, Subsets, summarizing functions. Data Cleaning, Finding and removing Duplicates, Sorting.

#### Module -3 (Statistics with R)

Analyzing Data, Summary statistics, Statistical Tests- Continuous Data, Discrete Data, Power tests, Common distributions- type arguments. Probability distributions, Normal distributions

#### Module -4(Data Visualization)

R Graphics- Overview, Customizing Charts, Graphical parameters, Basic Graphics functions, Lattice Graphics - Lattice functions, Customizing Lattice Graphics, Ggplot.

# Module - 5 (Regression Models)

Building linear models - model fitting, Predict values using models, Analyzing the fit, Refining the model, Regression- types, Unusual observation and corrective measures, Comparison of models, Generalized linear models - Logistic Regression, Poisson Regression, Nonlinear least squares

#### **Text Book**

1. Joseph Adler, "R in a Nutshell", Second edition, O'reilly, 2012

#### **Reference Books**

- 1. Jared P Lander, R for Everyone- Advanced analytics and graphics, Addison Wesley data analytics series, Pearson
- 2. Norman matloff, The art of R programming, A Tour of Statistical, Software Design, O'reilly
- 3. Robert Kabacoff, R in action, Data analysis and graphics with R, Manning
- 4. Garret Grolemund, Hands-on programming with R, Write your own functions and simulations, O'reilly

#### Sample Course Level Assessment Questions

#### **Course Outcome 1 (CO1):**

- 1. What is Coercion? How is it done in R?
- 2. Write a program to find the factorial of a number.
- 3. Write a program to compute roots of a quadratic equation.

#### Course Outcome 2 (CO2):

- 1. Write a program to read data from a table 'table123' in a database named 'db123' and display the values .
- 2. Explain Data cleaning in R
- 3. How missing data is handled in R?

#### Course Outcome 3(CO3):

- 1. Explain summary function in R
- 2. Illustrate how statistical testing is performed in R
- 3. Describe about probability distributions.

#### **Course Outcome 4 (CO4):**

1. Illustrate the use of ggplot() and various data visualization tools using appropriate datasets

#### **Course Outcome 5 (CO5):**

1. Illustrate the steps to predict the weight of a person when his height is unknown using linear regression for the data given below.

Height	151	174	138	186	128	136	179	163	152	130
Weight	63	81	56	91	47	57	76	72	62	48

# **Model Question paper**

**QP CODE:** 

Reg No:\_\_\_\_\_

Name :\_\_\_\_\_

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY FIFTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

#### Course Code: AIT 362

#### **Course Name: Programming in R**

#### Max.Marks:100

#### **Duration: 3 Hours**

# PART A

#### Answer all Questions. Each question carries 3 Marks

- 1. Write a R program to add element "23" to the vector (24,56,67) in the second position.
- 2. Discuss the general list operations in R with example.
- 3. Calculate the cumulative sum and cumulative product for the given data 23, 1, 7,2,8,10, 17 using R Program.
- 4. Explain aggregate function in R.
- 5. List the applications of R programming.

PAGES:3

- 6. Illustrate summary function.
- 7. List any three graphics functions.
- 8. Explain Lattice function.
- 9. Suppose that you have a dataset D1 and you design a linear regression model of degree 3 polynomial and you found that the training and testing error is "0" or in other terms it perfectly fits the data. What will happen when you fit a degree 2 polynomial in linear regression?
- 10. Explain logistic regression function in R.

# Part B

# Answer any one Question from each module. Each question carries 14 Marks

- 11.a Write a R program to extract every nth element from a vector. (7 marks)
- 11.b Find the Nth highest value of a vector in R. (7 marks)

OR

- 12.a Write a R program to create a data frame using two given vectors and display(7 marks) the duplicate elements and unique rows of the said data frame.
- 12.b Write a R program to compare two data frames to find the row(s) in the first(7 marks) data frame that are not present in the second data frame.
- 13.a Write a R program to call the (built-in) dataset air quality. Remove the(7 marks) variables 'Solar.R' and 'Wind' and display the data frame.
- 13.b Illustrate transformation functions in R.

(7 marks)

OR

14.a Write a R program to write the following data to a CSV file.

(7 marks)

	Country	Population_1_july_2018	Population_1_july_2019	change_in_percents
1	China	1,427,647,786	1,433,783,686	+0.43%
2	India	1,352,642,280	1,366,417,754	+1.02%
3	United States	327,096,265	329,064,917	+0.60%
4	Indonesia	267,670,543	270,625,568	+1.10%
5	Pakistan	212,228,286	216,565,318	+2.04%

(10x3=30)

- 14.b Given a file "auto.csv" of automobile data with the fields index, company,(7 marks) body-style, wheel-base, length, engine-type, num-of-cylinders, horsepower, average-mileage, and price, write R program to print total cars of all companies, Find the average mileage of all companies.
- 15.aWrite a note on data analysis using R.(7 marks)15.bExplain how statistical test are performed using R functions.(7 marks)OROR
- 16.a Write R code to generate the probability distribution table for number of(7 marks) successes from a binomial distribution where n=5 and probability of success in each trial is 0.25.
- 16.b Fit a Poisson distribution with the following data using the following data (7 marks)

X	0	1	2	3	4	5
F	142	156	69	27	5	1

#### OR

17 Given the sales information of a company as CSV file with the following, fields month\_number, face cream, facewash, toothpaste, bathingsoap, shampoo, moisturizer, total\_units, total\_profit. Write R codes to visualize the data as follows:

a) Toothpaste sales data of each month and show it using a scatter plot. (7 marks)b) Calculate total sale data for last year for each product and show it using a(7 marks)Pie chart.

#### OR

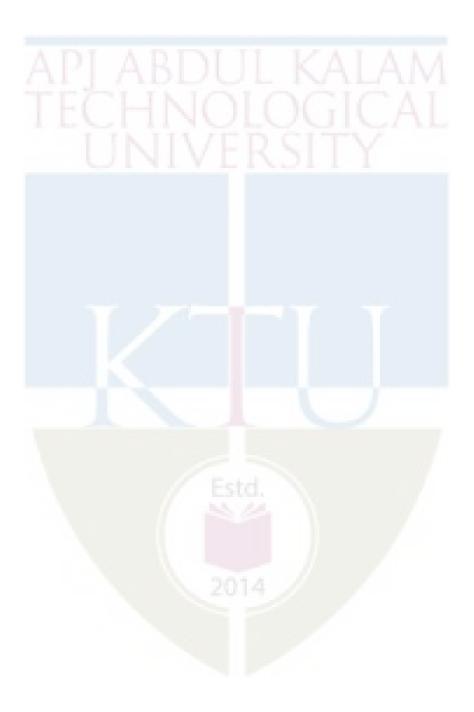
18.a	Explain ggplot() with and example.	(7 marks)					
18.b	18.b Describe how categorical data is visualized using R.						
19.a	Illustrate model fitting in simple linear model.	(7 marks)					
19.b	Explain different types of regression.	(7 marks)					
	OR						
20.a	Describe the unusual observations in the regression model.	(7 marks)					

20.b Explain corrective measures of unusual observations in regression modelling. (7 marks)

# **Teaching Plan**

No	Contents	No of Lecture Hours (35 Hours)
	Module -1 (Introduction to R)	(8 hours)
1.1	The R Environment- Command Line Interface and Batch processing, R Packages	1 hour
1.2	Variables, Data Types	1 hour
1.3	Vectors- vector operations and factor vectors	1 hour
1.4	List-List operations, Data Frames	1 hour
1.5	Matrices and arrays	1 hour
1.6	Control Statements- If and else, switch, if else	1 hour
1.7	Loops- For loops, While loops, Controlling loops	1 hour
1.8	Functions- Function as arguments, Named arguments	1 hour
	Module -2(Reading and writing data)	(8 hours)
2.1	Importing data from Text files and other software, Exporting data	1 hour
2.2	Importing data from databases- Database Connection packages	1 hour
2.3	Missing Data-NA, NULL	1 hour
2.4	Combining data sets, Transformations	1 hour
2.5	Binning Data, Subsets, summarizing functions	1 hour
2.6	Data Cleaning	1 hour
2.7	Finding and removing Duplicate	1 hour
2.8	Sorting	1 hour
	Module -3 (Statistics with R)	(6 hours)
3.1	Analyzing Data	1 hour
3.2	Summary statistics	1 hour
3.3	Statistical Tests- Continuous Data, Discrete Data, Power tests	1 hour
3.4	Common distributions- type arguments	1 hour
3.5	Probability distributions	1 hour
3.6	Normal distributions	1 hour
	Module -4(Data Visualization)	(6 hours)
4.1	R Graphics- Overview	1 hour
4.2	Customizing Charts	1 hour
4.3	Graphical parameters, Basic Graphics functions	1 hour
4.4	Lattice Graphics - Lattice functions	1 hour
4.5	Customizing Lattice Graphics	1 hour
4.6	ggplot	1 hour
	Module - 5 (Regression Models)	(7 hours)
5.1	Building linear models - model fitting	1 hour
5.2	Predict values using models, Analyzing the fit, Refining the model	1 hour
5.3	Regression- types of regression	1 hour

5.4	Unusual observations and corrective measures	1 hour
5.5	Comparison of models	1 hour
5.6	Generalized linear models -Logistic Regression, Poisson Regression	1 hour
5.7	Nonlinear least squares	1 hour



AMT372	Machine Learning Models and Storage	Category	L	T	Р	Credit	Year of Introduction
	Management	PEC	2	1	0	3	2020

# Preamble:

This course enables the learners to understand the basic machine learning models and different storage concepts. The course covers the standard and most popular supervised learning algorithms, storage technology, storage architecture, network storage system and securing and managing storage infrastructures. This course helps the students to choose the appropriate storage infrastructure for typical real world applications.

# **Prerequisite: Nil**

# Mapping of course outcomes with program outcomes

CO1	Illustrate the concepts of machine learning techniques and models(Cognitive Knowledge Level: Apply)				
CO2	Demonstrate various storage management technologies (Cognitive Knowledge Level: Apply)				
CO3	Explain Storage Systems Architecture and interaction of file systems (Cognitive Knowledge Level: Understand)				
CO4	Explain the different Network storage protocols (Cognitive Knowledge Level: Understand)				
CO5	Illustrate the concepts of management metric and standards(Cognitive Knowledge Level: Understand)				

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
C01	$\bigcirc$		$\bigcirc$		$\bigcirc$							
CO2	Ø	0	0	0.1		-						Ø
CO3	0	0	0	ι,β.	깃	1,	K	AI,	A.	Ŵ.		Ø
CO4	$\bigcirc$	0	0	-10	ιC.	ΣŢ	X	1	ĻΑ	L		Ø
CO5	Ø	Ø	0	11	VI	tΚ	51	1 Y				Ø

# Mapping of course outcomes with program outcomes

Abstract POs defined by National Board of Accreditation									
PO#	Broad PO	PO#	Broad PO						
PO1	Engineering Knowledge	<b>PO7</b>	Environment and Sustainability						
PO2	Problem Analysis	PO8	Ethics						
PO3	Design/Development of solutions	PO9	Individual and team work						
PO4	Conduct investigations of complex problems	PO10	Communication						
PO5	Modern tool usage	PO11	Project Management and Finance						
PO6	The Engineer and Society	PO12	Life long learning						

# **Assessment Pattern**

Bloom's	Continuo	ous Assessment Tests	End Semester	
Category	Test 1 (%)	Test 2 (%)	– Examination Marks (%)	
Remember	30	30	30	
Understand	30	30	30	
Apply	40	40	40	
Analyze				

Evaluate		
Create		

# **Mark Distribution**

Total Marks	CIE Marks	ESE Marks ESE D			
150	50	100	3		

#### **Continuous Internal Evaluation Pattern:**

Attendance				10 marks
Continuous A	Assessment Tests(Average of Int	err	nal Tests 1 & 2)	25 marks
Continuous A	Assessment Assignment			15 marks

# **Internal Examination Pattern**

Each of the two internal examinations has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed module), each with 7 marks. Out of the 7 questions, a student should answer any 5.

# **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 full questions from each module of which student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.

#### **Syllabus**

# Module – 1 (MACHINE LEARNING MODELS)

Introduction to Machine Learning, Examples of Machine Learning applications, Linear Regression: single & multiple variables, Classification: Logistic Regression - Decision

Trees, Overfitting & Underfitting, Bias -Variance trade-off, Support Vector Machines, Canonical Cases for Conditional Independence-Naive Bayes' Classifier.

# Module - 2(STORAGE TECHNOLOGY)

Information Storage-Data, Bigdata, Information, evolution of storage Architecture. Data Centre Infrastructure-Core elements, characteristics, Virtualization and Cloud Computing, Disk drive components, Physical disk structure, Zone Bit recording, Logical block addressing, Disk drive Performance, Direct Attached Storage, Storage design based on application requirements disk performance

# Module- 3(STORAGE SYSTEM ARCHITECTURE)

RAID, Implementation methods, RAID -Techniques-Striping, Mirroring, Parity. RAID Levels, RAID impact on disk performance. Components of an Intelligent Storage System-Front end, Cache, Back end, Storage provisioning-traditional vs virtual. Types of Intelligent storage systems

Backup and Archive- Backup Purpose, Backup Granularity, Backup methods , Backup architectures, Backup topologies

# Module - 4 (NETWORK STORAGE SYSTEM)

Fibre Channel Storage Area Networks- SAN and Its Evolution, Components of FC SAN, Fibre Channel Architecture, Fibre Channel Protocol Stack, FC SAN Topologies, Virtualization in SAN, IP SAN and FCoE- iSCSI- Components, FCIP Protocol Stack, Topology, FCoE.

Network-Attached Storage- Benefits of NAS, File Systems and Network File Sharing, Components of NAS, NAS Implementations-Unified NAS, Unified NAS Connectivity, Gateway NAS, Connectivity, NAS File-Sharing Protocols.

# Module - 5 (SECURING AND MANAGING STORAGE INFRACTURES)

Information Security Framework, Risk Triad, Storage Security Domains- Securing the Application Access Domain, Securing the Management Access Domain, Securing Backup, Replication, and Archive. Security Implementations in Storage Networking-FC SAN, NAS, IP SAN, Monitoring the Storage Infrastructure, Storage Infrastructure Management Activities, Storage Infrastructure Management Challenges, Information Lifecycle Management, Storage Tiering.

# **Text Books**

- 1. Introduction to machine learning, Second Edition, EthemAlpayd The MIT Press Cambridge, Massachusetts London, England
- 2. Information Storage and Management: Storing, Managing, and Protecting Digital Information in Classic, Virtualized, and Cloud Environments, Somasundaram, Gnanasundaram, Alok Shrivastava Editor: EMC Education Services, Wiley, 2012.

# ReferenceBooks

- 1. Information Storage and Management: Storing, Managing, and Protecting Digital Information, Antonio Cantiago, Wiley, 2009
- 2. Storage Area Network Essentials: A Complete Guide To Understanding And Implementing Sans, Richard Barker, Paul Massiglia, 2008
- 3. Storage Networks Explained: Basics and Application of Fibre Channel SAN, NAS, ISCSI, InfiniBand and FCoE, Ulf Troppens and Rainer Erkens, Wiley, 2009

# Sample Course Level Assessment Questions

# **Course Outcome 1 (CO1):**

- 1. Compare different machine learning paradigms with suitable examples.
- 2. Distinguish between overfitting and underfitting. How it can affect model generalization?

# **Course Outcome 2 (CO2):**

- 1. What is structured and unstructured data? Research the challenges of storing and managing unstructured data.
- 2. Discuss the benefits of information-centric storage architecture over server-centric storage architecture.

# **Course Outcome 3 (CO3):**

1. What is zoning? Discuss a scenario: a. Where WWN zoning is preferred over port zoning.

b. Where port zoning is preferred over WWN zoning.

- 2. Describe the process of assigning an FC address to a node when logging on to the network for the first time.
- 3. Seventeen switches, with 16 ports each, are connected in a full mesh topology. How many ports are available for host and storage connectivity?

# **Course Outcome 1 (CO4):**

- 1. SAN is configured for a backup-to-disk environment, and the storage configuration has additional capacity available. Can you have a NAS gateway configuration use this SAN-attached storage? Discuss the implications of sharing the backup-to-disk SAN environment with NAS.
- 2. Compared to a standard IP packet, what percentage of reduction can be realized in protocol overhead in an iSCSI, configured to use jumbo frames with an MTU value of 9,000 bytes?

# **Course Outcome 5 (CO5):**

 Describe Storage Management strategies for any two real world application scenarios (Storage Allocation to a New Server/Host, File System Space Management)

# **Model Question Paper**

# **QP CODE:**

Reg No: \_\_\_\_\_

Name: \_\_\_\_\_

PAGES:4

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

# SIXTH SEMESTER B TECH DEGREE EXAMINATION, MONTH & YEAR

# Course Code: AMT372

# **Course Name: Machine Learning Models and Storage Management**

# Max. Marks: 100

# **Duration: 3 Hours**

# PART A

# Answer All Questions. Each Question Carries 3 Marks

1.	Explain the significance of Naive assumption in Bayesian classifier	3
2.	Compare Classification with regression with an example	3
3.	What are the advantages of a virtualized data center over a classic data center?	3
4.	Which components constitute the disk service time? Which component contributes the largest percentage of the disk service time in a random I/O operation?	3
5.	What is meant by intelligent storage system.	3
6.	Why is RAID 1 not a substitute for a backup?	3
7.	Compare the topologies of FC-SAN, NAS, IP-SAN.	3
8.	What are the Factors affecting NAS performance.	3
9.	List the different security goals.	3
10.	How does the use of jumbo frames affect the NAS performance?	3 (10x3=30)

# Part B

# (Answer any one question from each module. Each question carries 14 Marks)

11. (a) For the following set of training samples, find which attribute can be chosen as the root for decision tree classification

Instance	Classification	a1	a2
1	DI /+ R D	Τ	T
2	1400	Т	Т
3	-()	Т	F
4	TICTIC	F	F
5	VINIV	F	Т
6		F	Т

(b) Explain the working principles of SVM classifiers.

# OR

- **12.** (a) What is overfitting? Explain Bias -Variance trade off.
  - (b) Use the following data to construct a linear regression model for the auto insurance premium as a function of number of years the vehicle used.

Years used	1	3	5	8	10	12
Insurance	9000	7000	6000	5000	4000	3000
Premium			/ F.	hts		

- 13. (a) The average I/O size of an application is 64 KB. The following specifications are available from the disk manufacturer: average seek time = 5 ms, 7,200 RPM, and transfer rate = 40 MB/s. Determine the maximum IOPS that could be performed with this disk for the application. Using this case as an example, explain the relationship between disk utilization and IOPS
  - (b) Illustrate any three Disk Drive Components.

OR

14. (a) Define the following terms(i)Disk Service Time(ii)Seek Time(iii)Rotational Latency

(8)

(7)

(6)

(7)

(7)

	(b) List the benefits and limitations of Direct Attached Storage	(6)
15.	(a) Explain the terms :Striping, Mirroring, Parity	(6)
	(b) Describe the Components of an Intelligent Storage System	(8)
	OR	
16.	(a) Explain the process of data recovery in case of a drive failure in RAID 5. What are the benefits of using RAID 3 in a backup application?	(7)
	(b) Explain the Array caching properties and algorithms.	(7)
17.	(a) Illustrate the NAS File-Sharing Protocols.	(10)
	(b) Explain Fibre Channel Architecture and Protocol Stack.	(4)
	OR	
18.	(a) Describe the Benefits of CAS?	(8)
	(b) Explain the Components of IP-SAN?	(6)
19.	(a) Explain how security is provided in application access domain and management access domain.	(10)
	(b) List out the challenges in storage Infrastructure management	(4)
	OR	
20.	(a) Describe the secure user access in NAS environment	(6)
	(b) Discuss different aspects of monitoring the storage infrastructure	(8)

# **Teaching Plan**

No	Contents						
	Module – 1 (FUNDAMENTALS) (7 hours)						
1.1	Introduction to Machine Learning, Examples of Machine Learning applications	1 hour					
1.2	Linear Regression: single & multiple variables,	1 hour					
1.3	Classification: Logistic Regression	1 hour					
1.4	Decision Trees	1 hour					
1.5	Overfitting & Underfitting, Bias Variance Trade-off	1 hour					
1.6	Support Vector Machines	1 hour					
1.7	Canonical Cases for Conditional Independence-Naive Bayes' Classifier.	1 hour					
	Module - 2(STORAGE TECHNOLOGY) (5 hours)						
2.1	Information Storage-Data, Bigdata, Information, evolution of storage Architecture	1 hour					
2.2	Data Centre Infrastructure-Core elements, characteristics, Virtualization and Cloud Computing						
2.3	Disk drive components, Physical disk structure, Zone Bit recording, Logical block addressing						
2.4	Disk drive Performance, Direct Attached Storage	1 hour					
2.5	Storage design based on application requirements disk performance	1 hour					
	Module - 3 (STORAGE SYSTEM ARCHITECTURE) (8 hours)						
3.1	RAID, Implementation methods, RAID -Techniques-Striping, Mirroring, Parity.						
3.2	RAID Levels, RAID impact on disk performance						
3.3	Components of an Intelligent Storage System-Front end, Cache, Back end,						
3.4	Storage provisioning-traditional vs virtual.						
3.5	Types of Intelligent storage systems						
3.6	Backup and Archive- Backup Purpose						

3.7	Backup Granularity, Backup methods, Backup architectures			
3.8	Backup topologies			
	Module - 4 (NETWORK STORAGE SYSTEM) (10 hours)			
4.1	Fibre Channel Storage Area Networks- SAN and Its Evolution, Components of FC SAN			
4.2	Fibre Channel Architecture, Fibre Channel Protocol Stack			
4.3	Zoning	1 hour		
4.4	FC SAN Topologies, Virtualization in SAN	1 hour		
4.5	IP SAN- FCoE and iSCSI, Components, topologies	1 hour		
4.6	iSCSI Protocol stack	1 hour		
4.7	FCoE, Components of FCoE	1 hour		
4.8	Network-Attached Storage- Benefits of NAS, File Systems and Network File Sharing,			
4.9	1.9 Components of NAS, NAS Implementations-Unified NAS, Unified NAS Connectivity, Gateway NAS, Connectivity,			
4.10	NAS File-Sharing Protocols			
	Module - 5 (MANAGING AND MONITORING) (7 hours)			
5.1	Managing & Monitoring: Management philosophies	1 hour		
5.2	Industry management standards (SNMP, SMI-S, CIM)	1 hour		
5.3	Standard framework applications, Key management metrics (thresholds, availability, capacity, security, performance)	1 hour		
5.4	Standard framework applications, Key management metrics (thresholds, availability, capacity, security, performance)	1 hour		
5.5	5 Provisioning & configuration change planning			
5.6	Problem reporting	1 hour		
5.7	prioritization and handling techniques, Management tools overview	1 hour		

		Category	L	Т	Р	Credit	Year of
AMT	COMPREHENSIVE						Introduction
308	COURSE WORK	PCC	1	0	0	1	2019

# **Preamble:**

The objective of this Course work is to ensure the comprehensive knowledge of each student in the most fundamental core courses in the curriculum. Five core courses credited from semesters 3, 4 and 5 are chosen for the detailed study in this course work. This course helps the learner to become competent in cracking GATE, placement tests and other competitive examinations

# **Prerequisite:**

- 1. Data Structures
- 2. Operating Systems
- **3. Introduction to Machine Learning**
- 4. Database Management Systems
- 5. Introduction to Artificial Intelligence

**Course Outcomes:** After the completion of the course the student will be able to

CO1:	Comprehend the concepts and applications of data structures (Cognitive Knowledge Level: <b>Understand</b> )				
CO2 :	Comprehend the concepts, functions and algorithms in operating system (Cognitive Knowledge Level: <b>Understand</b> ))				
CO3:	Comprehend the concepts of machine learning (Cognitive Knowledge Level: <b>Understand</b> )				
CO4:	Comprehend the fundamental principles of database design and manipulation (Cognitive Knowledge Level: Understand)				
CO5:	Comprehend the concepts of artificial intelligence (Cognitive Knowledge Level: <b>Understand</b> )				

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	$\bigcirc$	0	1		NI/	N	2	Xi	2	1 1		$\bigcirc$
CO2			1				2		3	1L		
CO3	$\bigcirc$	0	0	N	IV	11	0		Ι			$\bigcirc$
CO4	$\bigcirc$	Ø										$\bigcirc$
CO5		$\bigcirc$										$\bigcirc$

# Mapping of course outcomes with program outcomes

#### **Assessment Pattern**

Bloom's Category	End Semester Examination	
Remember	10	
Understand	20	
Apply	20	
Analyse	A A	
Evaluate		
Create		

# Mark distribution

Total Marks	CIE	ESE	ESE Duration		
50	0	50	1 hour		

**End Semester Examination Pattern:** Objective Questions with multiple choice, a maximum of four options. Question paper include fifty questions of one mark each, distributed equally from all the five identified courses.

# **Syllabus**

Full Syllabus of all five selected Courses.

- 1. Data Structures
- 2. Operating Systems
- 3. Introduction to Machine Learning
- 4. Database Management Systems
- 5. Introduction to Artificial Intelligence

#### **Course Contents and Lecture Schedule**

No	Торіс	No. of Lectures
1	DATA STRUCTURES	
1.1	Mock Test on Module 1, Module 2 and Module 3	1 hour
1.2	Mock Test on Module 4 and Module 5	1 hour
1.3	Feedback and Remedial class	
2	OPERATING SYSTEMS	
2.1	Mock Test on Module 1 and Module 2	1 hour
2.2	Mock Test on Module 3, Module 4 and Module 5	1 hour
2.3	Feedback and Remedial class	1 hour
3	INTRODUCTION TO MACHINE LEARNING	
3.1	Mock Test on Module 1, Module 2 and Module 3	1 hour
3.2	Mock Test on Module 4 and Module 5	1 hour
4	DATABASE MANAGEMENT SYSTEMS	
4.1	Mock Test on Module 1, Module 2 and Module 3	1 hour
4.2	Mock Test on Module 4 and Module 5	1 hour

4.3	Feedback and Remedial class			
5	INTRODUCTION TO ARTIFICIAL INTELLIGENCE			
5.1	Mock Test on Module 1, Module 2 and Module 3			
5.2	Mock Test on Module 4 and Module 5	1 hour		
5.3	Feedback and Remedial class	1 hour		

# **Model Question Paper**

**QP CODE:** 

Reg No: \_\_\_\_\_

Name:

PAGES:9

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY SIXTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

#### Course Code: AMT 308

#### **Course Name: Comprehensive Course Work**

Max. Marks: 50

**Duration: 1 Hour** 

Objective type questions with multiple choices. Mark one correct answer for each question. Each Question Carries 1 Mark

# A B-tree of order (degree)5 and of height 3 will have a minimum of \_\_\_\_\_ keys. A. 624

- B. 249
- C. 124
- D. 250
- 3. Construct a binary search tree by inserting 8, 6, 12, 3, 10, 9 one after another. To make the resulting tree as AVL tree which of the following is required?

- (A) One right rotation only
- (B) One left rotation followed by two right rotations
- (C) One left rotation and one right rotation
- (D) The resulting tree itself is AVL
- 4. In a complete 4-ary tree, every internal node has exactly 4 children or no child. The number of leaves in such a tree with 6 internal nodes is:
  - (A) 20 (B) 18 (C) 19 (D) 17
- 5. Select the postfix expression for the infix expression  $a+b-c+d^*(e/f)$ .
  - (A)  $ab+c-d+e^{f/}$  (B) ab+c-def/\*+
  - (C) abc-+def/\*+ (D) ab+c-def/\*+
- 6.

Consider a hash table of size seven, with

starting index zero, and a hash function (2x + 5)mod7. Assuming the hash table is initially empty, which of the following is the contents of the table when the sequence 1, 4, 9, 6 is inserted into the table using closed hashing? Note that '\_' denotes an empty location in the table.

(A) 9, \_, 1, 6, \_, \_, 4(B) 1, \_, 6, 9, \_, \_, 4(C) 4, \_, 9, 6, \_, \_, 1(D) 1, \_, 9, 6, \_, \_, 4

7. Compute the time complexity of the following function:

void function(int n)

ł

```
int count = 0;

for (int i=n/2; i<=n; i++)

for (int j=1; j<=n; j = j + 2)

for (int k=1; k<=n; k = k * 2)

count++;

}

A. O(n^2 \log n)

B. O(n \log^2 n)
```

- C.  $O(n^3)$
- D.  $O(n \log n^2)$
- How many distinct binary search trees can be created out of 6 distinct keys?
   (A) 7
   (B) 36
   (C) 140
   (D) 132
- 9. Which tree traversal performed on a binary search tree, results in ascending order listing of the keys?
  - A. Pre-order
  - B. In-order

C. Post-order

D. Level-order

10. You are given pointers to first and last nodes of a singly linked list, which of the following operations are dependent on the length of the linked list?

- (A) Delete the first element
- (B) Insert a new element as a first element
- (C) Add a new element at the end of the list
- (D) Delete the last element of the list

11. Suppose a disk has 400 cylinders, numbered from 0 to 399. At some time the disk arm is at cylinder 58, and there is a queue of disk access requests for cylinder 66, 349, 201, 110, 38, 84, 226, 70, 86. If Shortest-Seek Time First (SSTF) is being used for scheduling the disk access, the request for cylinder 86 is serviced after servicing \_\_\_\_\_ number of requests. (D)4

- (A) 1 (B) 2 (C)3

(D) 2KB

Calculate the internal fragmentation if

12. If frame size is 4KB then a paging system with page table entry of 2 bytes can address \_\_\_\_\_ bytes of physical memory. (A) 2^12 (B) 2<sup>16</sup> (C) 2<sup>18</sup> (D) 2<sup>2</sup>8

13.

page size is 4KB and process size is 103KB. (A) 3KB (B) 4KB (C) 1KB

14. Which of the following scheduling policy is likely to improve interactiveness?

(A) FCFS	(B) Round Robin
(C) Shortest Process Next	(D) Priority Based Scgeduling

15. Consider the following program

Semaphore X=1, Y=0		
Void A ( )	Voie	1 B ( )
{	{	
While (1)	2014	While (1)
{		{
P(X);		P(Y);
Print'1';		P(X);
V(Y);		Print'0';
}		V(X);
}		}
<i>,</i>	}	
	,	

The possible output of the program:

(A) Any number of 0's followed by any number of 1's.

(B) Any number of 1's followed by any number of 0's.

(C) 0 followed by deadlock

- (D) 1 followed by deadlock
- 16. In a system using single processor, a new process arrives at the rate of 12 processes per minute and each such process requires 5 seconds of service time. What is the percentage of CPU utilization?

(A) 41.66

(C) 240.00

(D) 60.00

- 17. A system has two processes and three identical resources. Each process needs two resources to proceed. Then
  - (A) Deadlock is possible (B) Deadlock is not possible
  - (C) Starvation may be present (D) Thrashing

(B) 100.00

- 18. Which of the following is true with regard to Round Robin scheduling technique?
  - (A) Responds poorly to short process with small time quantum.
  - (B) Works like SJF for larger time quantum
  - (C) Does not use a prior knowledge of burst times of processes.
  - (D) Ensure that the ready queue is always of the same size.
- 19. Thrashing can be avoided if
  - (A) the pages, belonging to working set of programs, are in main memory
  - (B) the speed of CPU is increased
  - (C) the speed of I/O processor is increased
  - (D) none of the above
- 20. The circular wait condition can be prevented by
  - (A) using thread
  - (B) defining a linear ordering of resource types
  - (C) using pipes
  - (D) all of the above
- 21. Application of machine learning methods to large databases is called

(A) Data Mining	(B) Artificial Intelligence

- (C) Big Data Computing (D) Internet of Things
- 22. If machine learning model output involves target variable, then that model is called as
  - (A) Descriptive Model (B) Predictive Model
  - (C) Reinforcement Learning (D) All of the above
- 23. In what type of learning labelled training data is used
  - (A) Unsupervised Learning (B) Supervised Learning
  - (C) Reinforcement Learning (D) Active Learning

24. In following type of feature selection method we start with empty feature set

(A) Forward Feature Selection	(B) Backward Feature Selection				
(C) Both A and B	(D) None of the above				
25. Which of the following is the best machine learning method?					
(A) Scalable	(B) Accuracy				
(C) Fast	(D) All of the above				
26. Data used to build a data mining model.					
(A) Training data	(B) Validation data				

(C) Test data (D) Hidden data

27. You are given reviews of few netflix series marked as positive, negative and neutral. Classifying reviews of a new netflix series is an example of

(A) Supervised learning	(B) Unsupervised learning				
(C) Semisupervised learning	g (D) Reinforcement learning				

28. Following are the types of supervised learning

(A) Classification	(B) Regression
(C) subgroup discovery	(D) all of the above

29. The output of training process in machine learning is

(A) machine learning model	(B) machine learning algorithm			
(C) null	(D) accuracy			

30. PCA is

(A) forward feature selection	(B) backward feature selection
(C) feature extraction	(D) all of the above

31. Let E1, E2 and E3 be three entities in an E/R diagram with simple single-valued attributes. R1 and R2 are two relationships between E1 and E2, where R1 is one-to-many, R2 is many-to-many. R3 is another relationship between E2 and E3 which is many-to-many. R1, R2 and R3 do not have any attributes of their own. What is the minimum number of tables required to represent this situation in the relational model?

```
(A) 3 (B) 4 (C) 5 (D) 6
```

- 32. Identify the minimal key for relational scheme R(U, V, W, X, Y, Z) with functional dependencies  $F = \{U \rightarrow V, V \rightarrow W, W \rightarrow X, VX \rightarrow Z\}$ (A) UV (B) UW (C) UX (D) UY
- 33. It is given that: "Every student need to register one course and each course registered by many students", what is the cardinality of the relation say "Register" from the "Student" entity to the "Course" entity in the ER diagram to implement the given requirement.

(A) M:1 relationship	(B) M:N relationship
(C) 1:1 relationship	(D) option (B) or(C)

34. Consider the relation branch( branch\_name, assets, branch\_city) SELECT DISTINCT T.branch\_name FROM branch T, branch S WHERE T.assets > L.assets

AND S.branch\_city = "TVM".

Finds the names of

- (A) All branches that have greater assets than all branches located in TVM.
- (B) All branches that have greater assets than some branch located in TVM.
- (C) The branch that has the greatest asset in TVM.
- (D) Any branch that has greater asset than any branch located in TVM.
- 35. Consider the following relation instance, where "A" is primary Key.

		0		,		1	2	2	
A1	A2	A3	A4						
1	1	1	Null						
5	2	5	1						
9	5	13	5						
13	13	9	15						
Which one	of th	e follo	wing car	n be a foreig	gn key tha	t refer	s to tl	ne same relati	ion?
(A) A2	(]	B) A3	(	C) A4	(D) AI	L			
(11) 112	()	<b>b</b> ) 115	(	C) III	(D)/11	<b></b>			

36. A relation R(ABC) is having the tuples(1,2,1),(1,2,2),(1,3,1) and (2,3,2). Which of the following functional dependencies holds well?

 $(A) A \to BC \quad (B) AC \to B \qquad (C) AB \to C \qquad (D) BC \to A$ 

- 37. Consider a relation R with attributes A, B, C, D and E and functional dependencies A→BC, BC→E, E→DA. What is the highest normal form that the relation satisfies?
  (A) BCNF
  (B) 3 NF
  (C) 2 NF
  (D) 1 NF
- 38. For the given schedule S, find out the conflict equivalent schedule. S:r1(x); r2(Z); r3(X); r1(Z); r2(Y); r3(Y);W1(X); W2(Z); W3(Y); W2(Y)
  - (A)  $T1 \rightarrow T2 \rightarrow T3$  (B)  $T2 \rightarrow T1 \rightarrow T3$

(C)  $T3 \rightarrow T1 \rightarrow T2$  (D) Not conflict serializable

39. Specialization is \_\_\_\_\_ process.

(A) top-down	(B) bottom up
(C) Both (A) and (B)	(D) none of these

40. If D1, D2, .., Dn are domains in a relational model, then the relation is a table, which is a

(B)  $D1 \times D2 \times ... \times Dn$ 

(D) D1–D2– ... –Dn

- subset of
  - (A) D1+D2+ ... +Dn

(C)  $D1 \cup D2 \cup \dots \cup Dn$ 

41. Artificial Intelligence is about\_\_\_\_

(A)Playing a game on Computer

(B)Making a machine Intelligent

- (C)Programming on Machine with your Own Intelligence
- (D)Putting your intelligence in Machine

42. Select the most appropriate situation for that a blind search can be used. (A)Real-life situation

- (B)Small Search Space(C)Complex game(D)All of the above
- 43. The application/applications of Artificial Intelligence is/are
  - (A)Expert Systems(B)Gaming(C)Vision Systems(D)All of the above
- 44. Among the given options, which search algorithm requires less memory?
  (A)Optimal Search
  (B)Depth First Search
  (C)Breadth-First Search
  (D)Linear Search
- 45. The component of an Expert system is \_\_\_\_\_

(A)Knowledge Base(B)Inference Engine(C)User Interface(D)All of the above

46. Which algorithm is used in the Game tree to make decisions of Win/Lose?
(A)Heuristic Search Algorithm
(B)DFS/BFS algorithm
(C)Greedy Search Algorithm
(D)Min/Max algorithm

47. Among the given options, which is not the required property of Knowledge representation? (A)Inferential Efficiency

(B)Inferential Adequacy

(C)Representational Verification

(D)Representational Adequacy

48. Which of the given language is not commonly used for AI?

(A)LISP(B)PROLOG(C)Python(D)Perl

49. A technique that was developed to determine whether a machine could or could not demonstrate

the artificial intelligence known as the\_\_\_\_ (A)Boolean Algebra (B)Turing Test (C)Logarithm (D)Algorithm

50. The available ways to solve a problem of state-space-search. (A)1 (B)2 (C)3 (D)4

QNo	Ans. Key								
1	(C)	11	(C)	21	(A)	31	(C)	41	(B)
2	(B)	12	(D)	22	(B)	32	(D)	42	(B)
3	(A)	13	(C)	23	(B)	33	(A)	43	(D)
4	(C)	14	(B)	24	(A)	34	(B)	44	(B)
5	(D)	15	(D)	25	(D)	35	(B)	45	(D)
6	(D)	16	(B)	26	(A)	36	(D)	46	(D)
7	(A)	17	(B)	27	(A)	37	(A)	47	(C)
8	(D)	18	(C)	28	(D)	38	(D)	48	(D)
9	(B)	19	(A)	29	(A)	39	(A)	49	(B)
10	(D)	20	(B)	30	(C)	40	(B)	50	(B)

AML332	Natural Language	CATEGORY	L	Т	Р	Credit
	Processing Lab	LAB	0	0	3	2

**Preamble**: The course should enable the students to provide theoretical concepts of language processing that shows how to explore interesting bodies of text. It helps to familiarize with fundamental topics in language processing that include tagging, classification and information extraction using Python programs. It covers the understanding of formal grammar to describe the structure of an unlimited set of sentences and also designing of existing corpora, the typical workflow for creating a corpus and the life cycle of a corpus

**Prerequisite**: Sound knowledge in Python programming.

Course Outcomes: After the completion of the course the student will be able to

CO#	Course Outcomes
CO1	Apply the concept of natural language processing (NLP) using Natural Language Toolkit (NLTK).( <b>Cognitive Knowledge Level: Apply</b> )
CO2	Build text corpora with tokenization, Stemming, Lemmatization and apply visualization techniques.( <b>Cognitive Knowledge Level: Apply</b> )
CO3	Evaluate the classifiers and choose the best classifier. (Cognitive Knowledge Level: Apply)
CO4	Create Artificial Intelligence applications for text data. (Cognitive Knowledge Level: Apply)

**Program Outcomes ( PO)** 

Mapping of course outcomes with program outcomes

	PO 1	PO 2	РО 3	РО 4	PO5	РО 6	РО 7	PO 8	PO 9	PO 10	PO 11	PO 12
C01			٢	A	BE	)U	L	Ķ	AL	A.	Ņ	
CO2	0		٢	iГ		VE/E	R	R ST		7		
CO3												
CO4	0		0	Ø	0							

Abstract POs defined by National Board of Accreditation								
PO#	Broad PO	PO#	Broad PO					
PO1	Engineering Knowledge	PO7	Environment and Sustainability					
PO2	Problem Analysis	PO8	Ethics					
PO3	Design/Development of solutions	PO9	Individual and teamwork					
PO4	Conduct investigations of complex problems	PO10	Communication					
PO5	Modern tool usage	PO11	Project Management and Finance					
PO6	The Engineer and Society	PO12	Lifelong learning					

#### **Assessment Pattern**

Bloom's Category	Continuous Assessment Tes (Internal Exam) Marks in percentage	St End Semester Examination Marks in percentage
Remember	ABL20UL K	20
Understand		20
Apply	60 – R S	60
Analyze		
Evaluate		
Create		

# **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	75	75	3 hours

# **Continuous Internal Evaluation Pattern:**

Attendance	: 15 marks
Continuous Evaluation in Lab	: 30 marks
Continuous Assessment Test	: 15 marks
Viva voce	: 15 marks

# **Internal Examination Pattern:**

The marks will be distributed as Algorithm 30 marks, Program 20 marks, Output 20 marks and Viva 30 marks. Total 100 marks which will be converted out of 15 while calculating Internal Evaluation marks.

#### **End Semester Examination Pattern:**

The marks will be distributed as Algorithm 30 marks, Program 20 marks, Output 20 marks and Viva 30 marks. Total 100 marks will be converted out of 75 for End Semester Examination.

Operating System to Use in Lab : Linux

Programming Language to Use in Lab : Python

# Fair Lab Record:

All the students attending the Natural language processing Lab should have a Fair Record. Every experiment conducted in the lab should be noted in the fair record. For every experiment, in the fair record, the right hand page should contain experiment heading, experiment number, date of experiment, aim of the experiment, procedure/algorithm followed, other such details of the experiment and final result. The left hand page should contain a print out of the respective code with sample input and corresponding output obtained. All the experiments noted in the fair record should be verified by the faculty regularly. The fair record, properly certified by the faculty, should be produced during the time of End Semester Examination for the verification by the examiners.

# **Syllabus**

1. Familiarize with Python natural language processing toolkit nltk.

- 2. Choose an English word, and see how it is used in the different example texts by making concordances.
- 3. Counting Vocabulary
  - 1. How many words (tokens) are there in the given text.
  - 2. How many different words (types) are there in the given text
  - 3. How many times does the word the occur in the text
  - 4. What is this as a percentage of all the words in the text?

4. Write Python program to perform preprocessing of text (Tokenization, Filtration, Script Validation, Stop Word Removal, Stemming).

- 5. Write a Python program to analyze text data using Constituency Parsing and Probabilistic Parsing.
- 6. Write a Python program to build Bag of Words model (BoW) in NLP.
- 7. Write a Python program to find the most similar sentence in the file to the given input sentence.
- 8. Implement Named Entity Recognition using nltk.

- 9. Write a Python program to find TF-IDF values of each words in a document.
- 10.Implement POS tagging using hidden markov model.
- 11. Implement a basic chatbot using Python.
- 12. Implement Machine Learning based Text Classification in Python.
- 13. Implement a chatbot using python.
- 14. Implement a language translator.

#### **Practice Questions**

- 1. Write a Python program to read a line of text, tokenize it and remove stopwords from it.
- 2. Write a Python program to tokenize a line of text, perform stemming and lemmatization with WordNet.
- 3. Write a Python program to replace words with its synonyms and negations with antonyms.
- 4. Write a Python program to replace words matching regular expressions.
- 5. Write a Python program to create a word list corpus.
- 6. Write a Python program to create a part-of-speech tagged word corpus after tokenizing a line of text,

filtering out stopwords, performing lemmatization and then performing part-of-speech tagging.

- 7. Write a Python program to tag proper names.
- 8. Write a Python program to perform tagging using regular expressions.
- 9. Write a Python program to perform classifier-based tagging.
- 10. Write a Python program to create NER tagged word corpus.
- 11. Write a Python program to rank the words in a document using TF-IDF.

#### **Reference Books:**

- Steven Bird, Evan Klein and Edward Loper, —Natural Language Processing with Pythonl, O'Reilly Media, Inc., 2009.
- 2. Multilingual natural Language Processing Applications: From Theory to Practice Daniel M. Bikel and Imed Zitouni, Pearson Publication
- 3. Natural Language Processing and Information Retrieval: Tanvier Siddiqui, U.S. Tiwary
- 4.. Python Text processing with NLTK 2.0 Cookbook , Jacob perkins, PACKT Publishing
- 5. Natural Languaage Processing with Python, Steven Bird, Ewan Klein, and Edward Loper, Oreilly.

AMD334	MINI PROJECT	CATEGORY	L	Т	Р	CREDITS
		PWS	0	0	3	2

# **Preamble:**

The objective of this course is to apply the fundamental concepts of Artificial Intelligence / Machine Learning principles for the effective development of an application/research project. Mini project enables the students to boost their skills, widen the horizon of thinking and their ability to resolve real life problems. The students are expected to design and develop a software/hardware project to innovatively solve a real-world problem.

**Prerequisite :**A sound knowledge in any programming language and Subjects studied up to sixth semester.

Course Outcomes: After the completion of the course the student will be able to

CO#	СО				
CO1	Identify technically and economically feasible problems of social relevance (Cognitive Knowledge Level: Apply)				
CO2	Identify and survey the relevant literature for getting exposed to related solutions (Cognitive Knowledge Level: Apply)				
CO3	Perform requirement analysis and identify design methodologies and develop adaptable and reusable solutions of minimal complexity by using modern tools and advanced programming techniques ( <b>Cognitive Knowledge Level: Apply</b> )				
CO4	Prepare technical report and deliver presentation(Cognitive Knowledge Level: Apply)				
CO5	Apply engineering and management principles to achieve the goal of the project (Cognitive Knowledge Level: Apply)				

#### PO5 **PO8** PO11 PO12 **PO1 PO2** PO3 **PO4 PO6 PO7 PO9** PO10 CO1 **CO2** CO3 **CO4 CO5**

	Abstract POs defined by National Board of Accreditation							
PO#	Broad PO	PO#	Broad PO					
PO1	Engineering Knowledge	<b>PO7</b>	Environment and Sustainability					
PO2	Problem Analysis	PO8	Ethics					
PO3	Design/Development of solutions	PO9	Individual and team work					
PO4	Conduct investigations of complex problems	PO10	Communication					
PO5	Modern tool usage	PO11	Project Management and Finance					
PO6	The Engineer and Society	PO12	Life long learning					

#### **Assessment Pattern**

#### **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	75	75	3

# **Split-up of Continuous Internal Evaluation :**

Attendance	10 marks
Project Guide	15 marks
Project Report	10 marks

Evaluation by the Committee (will be evaluating the level of completion and demonstration of functionality/specifications, presentation,

# Mapping of course outcomes with program outcomes

#### 40 marks

#### **Split-up of End Semester Examination:**

The marks will be distributed as

Presentation	: 30 marks
Demonstration : 20 r	narks DESTIL
Viva	: 25 marks.
Total	: 75 marks.
nga Dian	

# Course Plan

Student Groups with 3 or 4 members should identify a topic of interest in consultation with Faculty/Advisor. Review the literature and gather information pertaining to the chosen topic. State the objectives and develop a methodology to achieve the objectives. Carryout the design/fabrication or develop codes/programs to achieve the objectives. Innovative design concepts, performance, scalability, reliability considerations, aesthetics/ergonomic, user experience and security aspects taken care of in the project shall be given due weight.

The progress of the mini project is evaluated based on a minimum of two reviews. The review committee may be constituted with the Head of the Department or a senior faculty, Mini Project coordinator and project guide as the members. Innovative design concepts, reliability considerations, aesthetics/ergonomic aspects taken care of in the project shall be given due weight. The internal evaluation shall be made based on the progress/outcome of the project, reports and a viva-voce examination, conducted internally by a 3-member committee. A project report is required at the end of the semester. The product/application has to be demonstrated for its full design specifications.

#### **Guidelines for the Report preparation**

A bonafide report on mini project shall be submitted within one week after the final presentation. Minimum number of pages should be 40.

- Use Times New Roman font for the entire Report Chapter / Section Title Times New Roman 18, Bold; Heading 2 Times New Roman 16, Bold; Heading 3 Times New Roman 14, Bold; Body- Times New Roman12, Normal.
- Line Spacing Between Heading 2 3 lines, between lines in paragraph 1.5 lines.
- Alignments Chapter / Section Title Center, Heading 2 & 3 should be LeftAligned. Ensure that all body text is paragraph justified.
- Figures & Tables Ensure that all Figures and Tables are suitably numbered and given proper names/headings. Write figure title under the figure and table title above the table
- Suggestive order of documentation:
  - i. Top Cover
  - ii. Title page
  - iii. Certification page
  - iv. Acknowledgement
  - v. Abstract
  - vi. Table of Contents
  - vii. List of Figures and Tables
  - viii. Chapters
  - ix. Appendices, if any
  - x.

References/Bibliography



CST 382	INTRODUCTION TO SOFTWARE TESTING	Category	L	Т	Р	Credits	Year of Introduction
		VAC	3	1	0	4	2019

#### Preamble:

This is a course in theoretical computer science that includes test cases for white-box, blackbox, and grey-box approaches. This course describes the various techniques for test case design used to test software artifacts, including requirements, design, and code. The course includes different techniques for test case design based on graphs, programming language syntaxes and inputs. The course also covers symbolic execution using PEX tool.

Course Outcomes: After the completion of the course the student will be able to:-

	List a range of different software testing techniques and be able to apply specific unit
CO1	testing method to the projects using Junit. (Cognitive Knowledge Level:
	Understand)
	Explain mutation testing method for a given piece of code to identify hidden defects
CO2	that can't be detected using other testing methods. (Cognitive Knowledge Level:
	Understand)
CO3	Explain graph coverage criteria in terms of control flow graph and data flow graph
005	for a given program. (Cognitive Knowledge Level: Understand)
CO4	Demonstrate the importance of black-box approaches in terms of domain and
CO4	functional testing. (Cognitive Knowledge Level: Understand)
COF	Illustrate the use of PEX tool with symbolic execution. (Cognitive Knowledge
CO5	Level: Apply)

# Mapping of course outcomes with program outcomes

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO 9	PO10	PO11	PO12
C01	Ø		Ø									$\bigotimes$
CO2	Ø		Ø							Ø		$\bigcirc$

#### B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

CO3				$\bigcirc$					$\oslash$	$\oslash$
CO4	$\bigcirc$	$\bigotimes$		$\bigcirc$						
CO5	$\bigcirc$	0	0	0					٢	
APT ABDUL KALAM										

	Abstract POs defined by National Board of Accreditation							
PO#	Broad PO	PO#	Broad PO					
PO1	Engineering Knowledge	PO7	Environment and Sustainability					
PO2	Problem Analysis	PO8	Ethics					
PO3	Design/Development of solutions	PO9	Individual and team work					
PO4	Conduct investigations of complex problems	PO10	Communication					
PO5	Modern tool usage	PO11	Project Management and Finance					
PO6	The Engineer and Society	PO12	Life long learning					

#### **Assessment Pattern**

Bloom's Category	Continuous 2	Assessment Tests	End Semester Examination
	Test 1 (Marks)	Test 2 (Marks)	Marks
Remember	30	30	30
Understand	40	40	40
Apply	30	30	30
Analyze			
Evaluate			
Create			

Total	CIE	ESE	ESE
Marks	Marks	Marks	Duration
150	50	100	3 hours

#### **Mark Distribution**

<b>Continuous Internal Evaluation Pattern:</b>	
Attendance	: 10 marks
Continuous Assessment - Test	: 25 marks
Continuous Assessment - Assignment	: 15 marks

#### **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 questions (preferably, 3 questions each from the completed modules and 1 students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 students should answer all questions, a student should answer any 5.

#### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.

# **Syllabus**

# Module - 1 (Introduction to Software Testing)

Some Popular Errors – Ariane 5, Therac 25, Intel Pentium Bug. What is Software testing? Why should it be tested? Software Quality, Role of Testing. Testing Process - Level 0 thinking, Level 1 thinking, Level 2 thinking, Level 3 thinking, Level 4 thinking. Software Testing Terminologies - Verification, Validation and Testing, Faults, Error and Bug, Test cases, Coverage Criteria. Types of Testing- Unit testing, integration testing, System testing, Acceptance testing, Beta testing, Functional testing, Stress testing, Performance testing, Usability testing and Regression testing. Testing Methods - Black Box testing, White Box testing, Grey Box testing.

# Module - 2 (Unit Testing)

Concept of Unit testing. Static Unit testing. Dynamic Unit testing - Control Flow testing, Data Flow testing, Domain testing, Functional Program testing. Mutation testing - Mutation and Mutants, Mutation operators, Mutation score. Junit - Framework for Unit testing. Case Study - Mutation testing using Junit and Muclipse.

# Module - 3 (Unit Testing - White Box Approaches)

Overview of Graph Coverage Criteria. Structural Graph Coverage Criteria - Node/vertex coverage, Edge coverage, Edge pair coverage, Path coverage, Complete path coverage, Prime path coverage, Complete round trip coverage, Simple round trip coverage. Data Flow Criteria - du paths, du pairs. Subsumption Relationships among Graph Coverage Criteria. Graph Coverage for Source Code - Control flow graphs for code, CFG: If statement, CFG: If statement with return, CFG: Switch-case, CFG: Loops, CFG: Exceptions (try-catch). Example program – Statistics. Graph Coverage for Design Elements - Call graphs and classes, Class inheritance testing: Coverage criteria, Coverage criteria on inheritance graph, Data flow at the design level, Inter-procedural DU pairs, Coupling du-pairs example. Example - Quadratic Root. Case Study - Graph Based testing using JUnit Framework.

#### Module - 4 (Unit Testing - Black Box Approaches)

Domain Testing / Input Space Partitioning - Partitions of a set. Input domain modelling - Interface-based approach, Functionality-based approach. Identifying values. Multiple partitions of the inputdomain - All Combinations Coverage (ACoC), Each Choice Coverage (ECC), Pair-wise Coverage, T-wise Coverage, Base Choice Coverage, Multiple Base Choices Coverage. TriTyp example. Functional Testing - Functional Testing Concepts of Howden. Functional testing - Important Steps. Types of Functional testing - Equivalence Class Partitioning, Boundary Value Analysis, Decision Tables, Random Testing. Case Study - Black Box testing approaches using JUnit.

#### Module - 5 (Grey Box Testing Approaches)

Introduction to Grey Box testing - Why Grey Box testing, Gray Box Methodology, Advantages and Disadvantages. Techniques of Grey Box Testing - Matrix Testing, Regression Testing, Orthogonal Array Testing or OAT, Pattern Testing. An Introduction to PEX - Parameterized Unit Testing, The Testing Problem. Symbolic Execution – Example, Symbolic execution tree. PEX application Case Study – PEX.

#### **Text Books**

- 1. Paul Ammann and JeffOffutt ,Introduction to Software Testing.
- 2. KshirasagarNaik and PriyadarshiTripathy, Software Testing And Quality Assurance: Theory And Practice.

#### **Reference Materials**

- 1. <u>https://www.csc.ncsu.edu/academics/undergrad/honors/thesis/muclipsebinder.pdf</u> Muclipse tutorial.
- King, James C, "Symbolic Execution and Program Testing", Association for Computing Machinery, July 1976.
- 3.

# Sample Course Level Assessment Questions

Course Outcome 1 (CO1): Explain the following types of testing methods with examples.

- (i) Balck-box testing.
- (ii) White-box testing.
- (iii) Grey-box testing.

**Course Outcome 2 (CO2):** Define 12 mutants for the following method *power()* using effective mutation operators. Try to use each mutation operator at least once. Approximately, how many mutants do you think there would be, if all mutants for *power()* were created?

public static int power (int left, int right)

```
// Raises Left to the power of Right
```

- // precondition : Right >= 0
- // postcondition: Returns Left\*\*Right

```
intrslt;
```

rslt = Left;

```
if (Right == 0)
{
    rslt = 1;
}
else
{
    for (int i = 2; i <= Right; i++)
    rslt = rslt * Left;
}
return (rslt);
}</pre>
```

Course Outcome 3 (CO3): Draw the control flow graph and data flow graph of given piece of code.

public static double ReturnAverage(int value[],int AS, int MIN, int MAX){
/\*

Function: ReturnAverageComputes the average of all those numbers in the input array in the positive range [MIN, MAX]. The maximumsize of the array is AS. But, the array size could be smaller than AS in which case the endof input is represented by -999.

```
int i, ti, tv, sum;
doubleav;
i = 0; ti = 0; tv = 0; sum = 0;
while (ti< AS && value[i] != -999) {
ti++;
if (value[i] >= MIN && value[i] <= MAX) {
tv++;
sum = sum + value[i];
}
i++;
}
if (tv> 0)
av = (double)sum/tv;
```

\*/

else av = (double) -999; return (av); }

Course Outcome 4 (CO4): Explain the following with examples.

- 1. Input domain modelling.
- 2. All Combinations Coverage (ACoC)
- 3. Each Choice Coverage (ECC)
- 4. Pair-wise Coverage
- 5. T-wise Coverage
- 6. Base Choice Coverage
- 7. Multiple Base Choices Coverage.

**Course Outcome 5 (CO5):** Draw the symbolic execution tree for the following program code and explain the symbolic execution of testme ( $\alpha 1$ ,  $\alpha 2$ ).

```
1. int twice (int v) {
```

2. return 2 \* v;

```
3. }
```

- 4. void testme (int x, int y) {
- 5. z = twice (y);
- 6. if (z == x){
- 7. if (x > y + 10)
- 8. ERROR;
- 9. }
- 10. }
- 11. int main() {
- 12. x = sym input();
- 13. y = sym input();
- 14. testme ( x , y);
- 15. return(0);

16. }

# **Model Question Paper**

**QP CODE:** 

PAGES: 4

Reg No:\_\_\_\_

Name :\_\_\_\_\_

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

#### SIXTH SEMESTER B.TECH DEGREE EXAMINATION(MINOR), MONTH & YEAR

**Course Code: CST 382** 

**Course Name: Introduction to Software Testing** 

#### Max.Marks:100

**Duration: 3 Hours** 

#### PART A

#### Answer all Questions. Each question carries 3 Marks

- 1. Explain the differences between Validation and Verification.
- 2. Explain the differences between Fault, Error, and Bug?
- 3. Define Ground string, Mutation score, and Mutants.
- 4. What are the functions of Test driver and Test stubs in dynamic unit testing?
- 5. Define Node coverage, Edge coverage and Prime path coverage in a control flow graph.
- 6. What are du paths and du pairs in a data flow graph?
- 7. Explain the two approaches in input domain modelling.
- 8. Explain the difference between Equivalence Class Partitioning and Boundary Value Analysis.
- 9. Briefly explain three techniques of Grey box testing.
- 10. Explain the concept of symbolic execution with the help of a toy example.

(10x3=30)

#### Part B

#### (Answer any one question from each module. Each question carries 14 Marks)

11. (a) Explain the following types of testing(i) Black Box testing (ii) White Box testing (iii) Grey Box testing (14)

(iv) Unit testing (v) Integration testing (vi) System testing (vii) Acceptance testing

# OR

12.	(a)	<ul> <li>Explain the following coverage criterias based on the code fragment given</li> <li>below. (i) Functional coverage (ii) Statement coverage (iii)Conditional</li> <li>coverage (iv)Branch coverage</li> <li>int foo (int x, int y){</li> </ul>	(8)
		int $z = 0$ ; if $((x > 0) \&\& (y > 0))$ {	
		$z = x;$ } return z;	
		}	
	(b)	Write positive and negative test cases for an ATM Machine?	(6)
13.	(a)	Explain Dynamic unit test environment with a neat figure.	(8)
	(b)	Explain the major difference between control flow testing and data flow testing.	(6)
		OR	
14.		Explain seven types of mutation operators with neat examples.	(14)
15.	(a)	Explain touring, side trips and detours with a neat example.	(7)
	(b)	Explain simple path coverage and prime path coverage with the help of CFG given below.	(7)

OR

	(i) Simple <i>if</i> (ii) Simple <i>while</i> loop (iii) Simple <i>for</i> loop	(7)
(b)	Explain the following concepts with examples.	(7)
	(i) Call graph (ii) Inheritance graph (iii) Coupling du-pairs	
	(a) What are the four important steps in functional testing?	(7)
(b)	Briefly explain input domain modelling approaches.	(7)
		<ul><li>(b) Explain the following concepts with examples.</li><li>(i) Call graph (ii) Inheritance graph (iii) Coupling du-pairs</li><li>(a) What are the four important steps in functional testing?</li></ul>

		OR	
18.	(a)	Consider the triangle classification program with a specification:	(6)
		The program reads floating values from the standard input. The three values	
		A, B, and C are interpreted as representing the lengths of the sides of	
		triangle. The program then prints a message to the standard output that states	
		whether the triangle, if it can be formed, is scalene, isosceles, equilateral, or	
		right angled. Determine the following for the above program:	
		(i) For the boundary condition $A + B > C$ case (scalene triangle),	
		identify test cases to verify the boundary.	
		(ii) For the boundary condition $A = C$ case (isosceles triangle), identify	
		test cases to verify the boundary.	
		(iii) For the boundary condition $A = B = C$ case (equilateral triangle),	
		identify test cases to verify the boundary.	
	(b)	Develop a decision table to generate test cases for this specification.	(8)
19.	(a)	Explain the importance of grey box testing, its advantages and disadvantages.	(9)
	(b)	Explain the concept of symbolic execution tree.	(5)
		OR	
20.	(a)	Consider the code fragment given below: -	(7)
		1. POWER: PROCEDURE(X, Y); 2. $Z \leftarrow 1$ ;	
		3. $J \leftarrow 1;$	

4. LAB: IF  $Y \ge J$  THEN

- 5. DO;  $Z \leftarrow Z * X$ ;
- 6.  $J \leftarrow J + 1;$
- 7. GO TO LAB; END;
- 8. RETURN (Z);
- 9. END;

**a**) Explain Symbolic execution of POWER ( $\alpha$ l,  $\alpha$ 2).

(b) Explain Execution tree for POWER ( $\alpha$ l,  $\alpha$ 2) in the above code fragment.

7)

# **TEACHING PLAN**

Index	Topics	No. of Hours (45)
	Module 1 (Introduction to Software Testing) 9 Hours	
1.1	Some Popular Errors– Ariane 5, Therac 25, Intel Pentium Bug.	1 Hour
1.2	What is Software testing? Why should it be tested? Software Quality, Role of Testing.	1 Hour
1.3	Testing Process - Level 0 thinking, Level 1 thinking, Level 2 thinking, Level 3 thinking, Level 4 thinking.	1 Hour
1.4	Software Testing Terminologies- Verification, Validation and Testing, Faults, Error and Bug, Test cases, Coverage Criteria.	1 Hour
1.5	Types of Testing- Unit testing, integration testing, System testing, Acceptance testing, Beta testing	1 Hour
1.6	Functional testing, Stress testing	1 Hour
1.7	Performance testing, Usability testing and Regression testing.	1 Hour
1.8	Testing Methods - Black Box testing	1 Hour
1.9	Grey Box testing.	1 Hour
	Module 2 (Unit testing) 8 Hours	

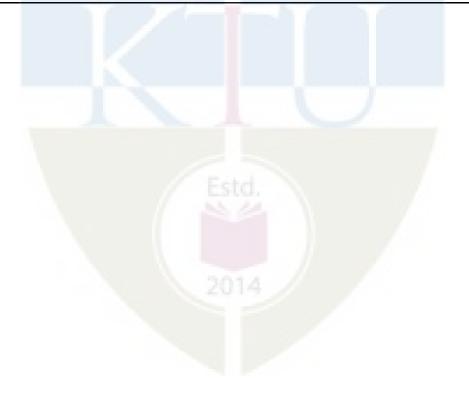
# B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

2.1	Concept of Unit testing.	1 Hour
2.2	Static Unit testing.	1 Hour
2.3	Dynamic Unit testing - Control Flow testing, Data Flow testing	1 Hour
2.4	Domain testing, Functional Program testing.	
2.5	Mutation testing - Mutation and Mutants, Mutation operators, Mutation score.	1 Hour
2.6	Junit - Framework for Unit testing.	1 Hour
2.7	Case Study - Mutation testing using Junit	1 Hour
2.8	Case Study - Mutation testing using Muclipse	1 Hour
	Module 3 (Unit Testing:- White Box Approaches) 10 Hours	
3.1	Overview of Graph Coverage Criteria	1 Hour
3.2	Structural Graph Coverage Criteria - Node/vertex coverage, Edge coverage, Edge pair coverage, Path coverage	1 Hour
3.3	Complete path coverage, Prime path coverage, Complete round trip coverage, Simple round trip coverage.	1 Hour
3.4	Data Flow Criteria - du paths, du pairs	1 Hour
3.5	Subsumption Relationships among Graph Coverage Criteria.	1 Hour
3.6	Graph Coverage for Source Code - Control Flow Graphs (CFG) for code, CFG: If statement, CFG: If statement with return, CFG: Switch-case, CFG: Loops, CFG: Exceptions (try-catch). Example program - Statistics	1 Hour
3.7	Graph Coverage for Design Elements - Call graphs and classes, Class inheritance testing: Coverage criteria, Coverage criteria on inheritance graph,	1 Hour

3.8	Data flow at the design level, Inter-procedural DU pairs, Coupling du-pairs example. Example - Quadratic Root	1 Hour
3.9	Case Study - Graph Based testing using JUnit Framework. (Lecture 1)	1 Hour
3.10	Case Study - Graph Based testing using JUnit Framework. (Lecture 2)	1 Hour
	Module 4 (Unit Testing: - Black Box Approaches) 9 Hours	
4.1	Domain Testing / Input Space Partitioning - Partitions of a set.	1 Hour
4.2	Input domain modelling - Interface-based approach, Functionality-based approach.	1 Hour
4.3	Identifying values.	1 Hour
4.4	Multiple partitions of the input domain - All Combinations Coverage (ACoC), Each Choice Coverage (ECC), Pair-wise Coverage, T-wise Coverage, Base Choice Coverage, Multiple Base Choices Coverage.	1 Hour
4.5	TriTyp example.	1 Hour
4.6	Functional Testing - Functional Testing Concepts of Howden. Important Steps.	1 Hour
4.7	Types of Functional testing - Equivalence Class Partitioning, Boundary Value Analysis	1 Hour
4.8	Decision Tables, Random Testing.	1 Hour
4.9	Case Study - Black Box testing approaches using JUnit.	1 Hour
	Module 5 (Grey Box Testing Approaches) 9 Hours	
5.1	Introduction to Grey Box testing - Why Grey Box testing, Gray Box Methodology, Advantages and Disadvantages.	1 Hour
5.2	Techniques of Grey Box Testing - Matrix Testing, Regression Testing,	1 Hour

# B TECH ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING

5.3	Orthogonal Array Testing or OAT, Pattern Testing.					
5.4	An Introduction to Pex - Parameterized Unit Testing, The Testing Problem.					
5.5	Symbolic Execution – Example, Symbolic execution tree.	1 Hour				
5.6	PEX application.	1 hour				
5.7	Case Study – PEX (Lecture 1)	1 Hour				
5.8	Case Study – PEX (Lecture 2)	1 Hour				
5.9	Case Study – PEX (Lecture 3)	1 Hour				



CST 384	CONCEPTS IN DEEP	Category	L	Т	Р	Credits	Year of Introduction
	LEARNING	VAC	3	1	0	A 4 A	2019

#### **Preamble**:

This course aims to introduce the learner to an overview of the concepts and algorithms involved in deep learning. Deep learning is a subfield of machine learning, a subfield of artificial intelligence. Basic concepts and application areas of machine learning, deep networks, convolutional neural network and recurrent neural network are covered here. This is a foundational program that will help students understand the capabilities, challenges, and consequences of deep learning and prepare them to participate in the development of leading-edge AI technology. They will be able to gain the knowledge needed to take a definitive step in the world of AI.

CO1	Demonstrate basic concepts in machine learning.(Cognitive Knowledge Level: Understand)
CO2	Illustrate the validation process of machine learning models using hyper-parameters and validation sets. (Cognitive Knowledge Level: Understand)
CO3	Demonstrate the concept of the feed forward neural network and its training process. (Cognitive Knowledge Level: Apply)
CO4	Build CNN and Recurrent Neural Network (RNN) models for different use cases. (Cognitive Knowledge Level: Apply)
CO5	Use different neural network/deep learning models for practical applications. (Cognitive Knowledge Level: Apply)

**Prerequisite:** Sound knowledge in Basics of linear algebra and probability theory.

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	$\bigcirc$	$\oslash$	$\bigcirc$	$\bigcirc$	DT	NI I		Z	A T	Αλ	A	$\oslash$
CO2	$\bigcirc$	$\oslash$	$\oslash$	$\bigcirc$	DL	ñ	10	N			1	$\oslash$
CO3	$\bigcirc$	$\bigcirc$	$\oslash$	$\oslash$	$\oslash$	XF	R		ΓV	n.		$\oslash$
CO4	$\bigcirc$	$\oslash$	$\oslash$	$\oslash$	$\oslash$	$\bigcirc$	1.					$\oslash$
CO5	$\bigcirc$	$\oslash$	$\oslash$	$\bigcirc$	$\oslash$	$\oslash$						$\oslash$

	Abstract POs defined by National Board of Accreditation								
PO#	Broad PO	PO#	Broad PO						
PO1	Engineering Knowledge	PO7	Environment and Sustainability						
PO2	Problem Analysis	PO8	Ethics						
PO3	Design/Development of solutions	Individual and team work							
PO4	Conduct investigations of complex problems	PO10	Communication						
PO5	Modern tool usage	PO11	Project Management and Finance						
PO6	The Engineer and Society	PO12	Life long learning						

Bloom's Category	Continuous Assessm	End Semester Examination		
	Test1 (Percentage)	Test2 (Percentage)	Marks	
Remember	30	30	30	
Understand	40	40	40	
Apply	<u>30</u>	- 30	30	
Analyse				
Evaluate				
Create				

#### **Assessment Pattern**

# **Mark Distribution**

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3 hours

#### **Continuous Internal Evaluation Pattern:**

Attendance : 10 marks

Continuous Assessment Tests : 25 marks

Continuous Assessment Assignment : 15 marks

# **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. First Internal Examination shall be preferably conducted after completing the first half of the syllabus and the Second Internal Examination shall be preferably conducted after completing remaining part of the syllabus.

There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly covered module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly covered module), each with 7 marks. Out of the 7 questions in Part B, a student should answer any 5.

### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have maximum 2 sub-divisions and carry 14 marks.

#### **Syllabus**

# **INTRODUCTION TO DEEP LEARNING**

(General Instructions: Instructors are to introduce students to any one software platform and demonstrate the working of the algorithms in the syllabus using suitable use cases and public datasets to give a better understanding of the concepts discussed. Tutorial hour may be used for this purpose)

# **Module-1** (Introduction)

Key components - Data, models, objective functions, optimization algorithms, Learning algorithm. Supervised learning- regression, classification, tagging, web search, page ranking, recommender systems, sequence learning, Unsupervised learning, Reinforcement learning, Historical Trends in Deep Learning. Other Concepts - overfitting, underfitting, hyperparameters and validation sets, estimators, bias and variance.

# Module- 2 (Optimization and Neural Networks)

Neural Networks –Perceptron, Gradient Descent solution for Perceptron, Multilayer perceptron, activation functions, architecture design, chain rule, back propagation, gradient based learning. Introduction to optimization– Gradient based optimization, linear least squares. Stochastic gradient descent, Building ML algorithms and challenges.

#### Module -3 (Convolutional Neural Network)

Convolutional Neural Networks – convolution operation, motivation, pooling, Structure of CNN, Convolution and Pooling as an infinitely strong prior, variants of convolution functions, structured outputs, data types, efficient convolution algorithms. Practical challenges of common deep learning architectures- early stopping, parameter sharing, dropout. Case study: AlexNet, VGG, ResNet.

### Module- 4 (Recurrent Neural Network)

Recurrent neural networks – Computational graphs, RNN design, encoder – decoder sequence to sequence architectures, deep recurrent networks, recursive neural networks, modern RNNs LSTM and GRU, Practical use cases for RNNs.

#### **Module-5** (Application Areas)

Applications – computer vision, speech recognition, natural language processing, common word embedding: continuous Bag-of-Words, Word2Vec, global vectors for word representation (GloVe). Research Areas – autoencoders, representation learning, boltzmann machines, deep belief networks.

#### **Text Book**

- 1. Ian Goodfellow, YoshuaBengio, Aaron Courville, Deep Learning, MIT Press 2015 ed.
- 2. Aston Zhang, Zachary C. Lipton, Mu Li, and Alexander J. Smola, Dive into Deep Learning, August 2019.
- 3. Neural Networks and Deep Learning, Aggarwal, Charu C., c Springer International Publishing AG, part of Springer Nature 2018

# **Reference Books**

- 1. Neural Smithing: Supervised Learning in Feedforward Artificial Neural Networks by Russell Reed, Robert J MarksII, A Bradford Book,2014
- 2. Practical Convolutional Neural Networks by MohitSewak, Md. Rezaul Karim, PradeepPujari,Packt Publishing 2018
- 3. Hands-On Deep Learning Algorithms with Python by SudharsanRavichandran,Packt Publishing 2019
- 4. Deep Learning with Python by Francois Chollet, Manning Publications Co., 2018

#### Sample Course Level Assessment Questions

### **Course Outcome 1 (CO1):**

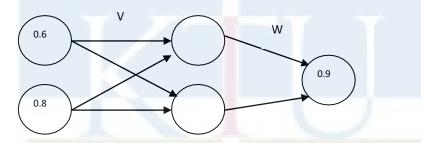
- 1. Compare regression and classification.
- 2. Define supervised learning? Distinguish between regression and classification.
- 3. Discuss the different learning approaches used in machine learning.

#### Course Outcome 2 (CO2):

- 1. What are hyperparameters? Why are they needed?
- 2. What issues are to be considered while selecting a model for applying machine learning in a given problem?

#### Course Outcome 3 (CO3):

1. Update the parameters  $V_{11}$  in the given MLP using back propagation with learning rate as 0.5 and activation function as sigmoid. Initial weights are given as  $V_{11}$ = 0.2,  $V_{12}$ =0.1,  $V_{21}$ =0.1,  $V_{22}$ =0.3,  $V_{11}$ =0.2,  $W_{11}$ =0.5,  $W_{21}$ =0.2



- 2. Draw the architecture of a multi-layer perceptron.
- 3. Derive update rules for parameters in the multi-layer neural network through the gradient descent.

#### **Course Outcome 4 (CO4):**

- Give two benefits of using convolutional layers instead of fully connected ones for visual tasks.
- 2. Suppose that a CNN was trained to classify images into different categories. It performed well on a validation set that was taken from the same source as the training set but not on a testing set. What could be the problem with the training of such a CNN? How will you ascertain the problem? How can those problems be solved?
- 3. Explain how the cell state is updated in the LSTM model from Ct-1 to Ct
- 4. Show the steps involved in an LSTM to predict stock prices.

#### **Course Outcome 5 (CO5):**

- 1. Explain how the cell state is updated in the LSTM model from Ct-1 to Ct
- 2. Show the steps involved in an LSTM to predict stock prices.
- 3. Illustrate the workings of the RNN with an example of a single sequence defined on a vocabulary of four words.

#### **Course Outcome 6 (CO6):**

- 1. Development a deep learning solution for problems in the domain i) natural language processing or ii Computer vision (Assignment
- 2. Illustrate the workings of the RNN with an example of a single sequence defined on a vocabulary of four words.

<b>Model Question Pape</b>	r
----------------------------	---

QP CODE:

PAGES:4

Reg	No:	

Name:

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY SIXTH SEMESTER B.TECH DEGREE EXAMINATION(MINOR), MONTH & YEAR

#### Course Code: CST 384

#### **Course Name: CONCEPTS IN DEEP LEARNING**

Max. Marks:100

#### **Duration: 3 Hours**

#### PART A

#### Answer all Questions. Each question carries 3 Marks

- 1. Distinguish between supervised learning and Reinforcement learning. Illustrate with an example.
- 2. Differentiate classification and regression.
- 3. Compare overfitting and underfitting. How it can affect model generalization.

- 4. Why does a single perceptron cannot simulate simple XOR function? Explainhow this limitation is overcome?
- 5. Illustrate the strengths and weaknesses of convolutional neural networks.
- 6. Illustrate convolution and pooling operation with an example
- 7. How many parameters are there in AlexNet? Why the dataset size (1.2 million) is important for the success of AlexNet?
- 8. Explain your understanding of unfolding a recursive or recurrent computation into a computational graph.
- 9. Illustrate the use of deep learning concepts in Speech Recognition.
- 10. What is an autoencoder? Give one application of an autoencoder

(10x3=30)

#### Part B)

# (Answer any one question from each module. Each question carries 14 Marks)

11.	(a)	"A computer program is said to learn from experience E with respect to some	
		class of	(10)
		tasks T and performance measure P, if its performance at tasks in T, as measured by P, improves with experience E." What is your understanding of the terms task, performance and experience. Explain with two example	
	(b)	"How does bias and variance trade-off affect machine learning algorithms?	(4)

#### OR

12.	(a)	Illustrate the concepts of Web search, Page Ranking, Recommender systems	
		with suitable examples.	(10)
	(b)	List and discuss the different hyper parameters used in fine tuning the	(4)

(7)

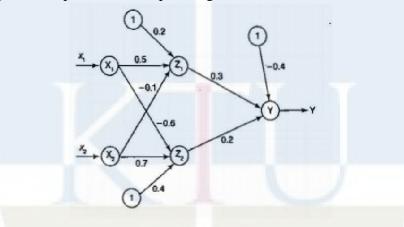
(7)

traditional machine learning models

- 13. (a) How multilayer neural networks learn and encode higher level features from input features.
  - (b) Explain gradient decent and delta rule? Why stochastic approximation to gradient descent is needed?

#### OR

14. (a) Find the new weights for the network using backpropogation algorithm, the network is given with a input pattern[-1,1] and target output as +1, Use learning rate of alpha=0.3 and bipolar sigmoid function.



(b) Write an algorithm for backpropgation which uses stochastic gradient descent (7) method. Comment on the effect of adding momentum to the network.

15.	(a) Input to CNN architecture is a color image of size 112x112x3. The first	(5)
	convolution layer comprises of 64 kernels of size 5x5 applied with a stride	
	of 2 and padding 0. What will be the number of parameters?	

- (b) Let X=[-1, 0, 3, 5] W=[.3, .5, .2, .1] be the the input of i<sup>th</sup> layer of a neural network and to apply softmax function. What should be the output of it?
- (c) Draw and explain the architecture of convolutional network (5)

#### OR

16.	(a)	Explain the	concept behind	i) Early stopping	ng ii) dropout iii	) weight decay	(9)
-----	-----	-------------	----------------	-------------------	--------------------	----------------	-----

	(b)	How backpropagation is used to learn higher-order features in a convolutional Network?	(5)
17.	(a)	Explain the working of RNN and discuss how backpropagation through time is used in recurrent networks.	(8)
	(b)	Describe the working of a long short term memory in RNNs.	(6)
		OR	
18.	(a)	What is the vanishing gradient problem and exploding gradient problem?	(8)
	(b)	Why do RNNs have a tendency to suffer from exploding/vanishing gradient? How to overcome this challenge?	(6)
19.	(a)	Explain any two word embedding techniques	(8)
	(b)	Explain the merits and demerits of using Auto encoders in Computer Vision.	(6)
		OR	
20.	(a)	Illustrate the use of representation learning in object classification.	(7)
	(b)	Compare Boltzmann Machine with Deep Belief Network.	(7)

# **Teaching Plan**

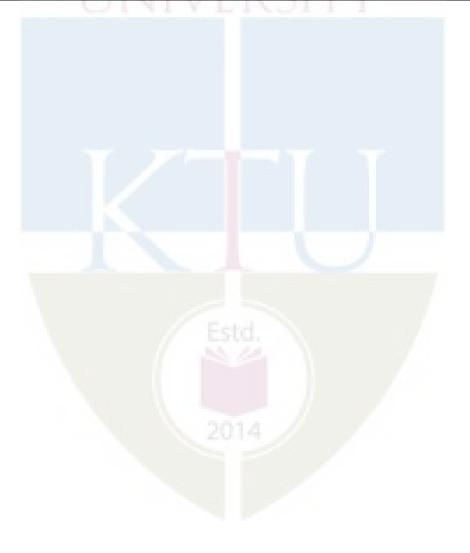
	<b>CONCEPTS IN DEEP LEARNING (45 Hours)</b>			
	Module 1 : Introduction (9 hours)			
1.1	Key components - Data, models, objective functions, optimization algorithms. (TB2: Section 1.1-1.2)	1 hour		

1.2	Learning algorithm (TB1: Section 5.1), Supervised learning- regression, classification (TB2: Section 1.3.1)	1 hour
1.3	tagging, web search, page ranking (TB2: Section 1.3.1)	1 hour
1.4	Recommender systems, Sequence learning, Unsupervised learning, Reinforcement learning(TB2: Section 1.3.2-1.3.4)	1 hour
1.5	Historical Trends in Deep Learning (TB1: Section 1.2).	1 hour
1.6	Concepts: over-fitting, under-fitting, hyperparameters and validation sets. (TB1: Section 5.2-5.3)	1 hour
1.7	Concepts: Estimators, bias and variance. (TB1: Section 5.4)	1 hour
1.8	Demonstrate the concepts of supervised learning algorithms using a suitable platform.	1 hour
1.9	Demonstrate the concepts of unsupervised using a suitable platform.	1 hour
	Module 2 : Optimization and Neural Networks (9 hours)	
2.1	Perceptron, Stochastic Gradient descent, Gradient descent solution for perceptron (TB3: Section 1.1 - 1.2.1)	1 hour
2.2	Multilayer perceptron (TB3: Section 1.2.2), (TB1: Section 6.1,6.3)	1 hour
2.3	Activation functions- Sigmoid, tanh, Softmax, ReLU, leaky ReLU (TB3: Section 1.2.1.3 - 1.2.1.5)	1 hour
2.4	Architecture design (TB1: Section 6.4, TB3: Section 1.6)	1 hour
2.5	Chain rule, back propagation (TB3: Section 1.3)	1 hour

2.6	Gradient based learning (TB1: Section 6.2)	1 hour
2.7	Gradient based optimization (TB1: Section 4.3)	1 hour
2.8	Linear least squares using a suitable platform. (TB1: Section 4.5)	1 hour
2.9	Building ML Algorithms and Challenges (TB3: 1.4, TB1: 5.10-5.11)	1 hour
	Module 3 :Convolution Neural Network (10 hours)	
3.1	Convolution operation, Motivation, pooling (TB1:Section 9.1-9.3)	1 hour
3.2	Structure of CNN (TB3: Section 8.2)	1 hour
3.3	Convolution and Pooling as an infinitely strong prior (TB1: Section 9.4)	1 hour
3.4	Variants of convolution functions – multilayer convolutional network, tensors, kernel flipping, downsampling, strides and zero padding. (TB1: Section 9.5)	1 hour
3.5	Variants of convolution functions - unshared convolutions, tiled convolution, training different networks. (TB1: Section 9.5)	1 hour
3.6	Structured outputs, data types (TB1: Section 9.6-9.7)	1 hour
3.7	Efficient convolution algorithms. (TB1: Section 9.8,9.10)	1 hour
3.8	Practical challenges of common deep learning architectures- early Stopping (TB3: 4.6)	1 hour
3.9	Practical challenges of common deep learning architectures- parameter sharing, drop-out (TB3: Section 4.9, 4.5.4)	1 hour
3.10	Case Study: AlexNet,VGG, ResNet. (TB3: Section 8.4.1-8.4.3,8.4.5)	1 hour

	Module 4 : Recurrent Neural Network (8 hours)	
4.1	Computational graphs (TB1: Section 10.1)	1 hour
4.2	RNN (TB1: Section 10.2-10.3)	1 hour
4.3	Encoder – decoder sequence to sequence architectures. (TB1: Section 10.4)	1 hour
4.4	Deep recurrent networks (TB1: Section 10.5)	1 hour
4.5	Recursive neural networks, Modern RNNs, (TB1: Section 10.6, 10.10)	1 hour
4.6	LSTM and GRU (TB1: Section 10.10, TB3: Section 7.5-7.6)	1 hour
4.7	Practical use cases for RNNs. (TB1: Section 11.1-11.4)	1 hour
4.8	Demonstrate the concepts of RNN using a suitable platform.	1 hour
	Module 5 : Applications and Research (9 hours)	
5.1	Computer vision. (TB1: Section 12.2)	1 hour
5.2	Speech recognition. (TB1: Section 12.3)	1 hour
5.3	Natural language processing. (TB1: Section 12.4)	1 hour
5.4	Common Word Embedding -: Continuous Bag-of-Words, Word2Vec (TB3: Section 2.6)	1 hour
5.5	Common Word Embedding -: Global Vectors for Word Representation(GloVe) (TB3: Section 2.9.1- Pennigton 2014)	1 hour
5.6	Brief introduction on current research areas- Autoencoders, Representation learning. (TB3: Section 4.10)	1 hour

5.7	Brief introduction on current research areas- representation learning. (TB3: Section 9.3)	1 hour
5.8	Brief introduction on current research areas- Boltzmann Machines, Deep belief networks. (TB1: Section 20.1, TB3 Section 6.3)	1 hour
5.9	Brief introduction on current research areas- Deep belief networks. (TB1: Section 20.3)	1 hour



CS'	I NETWORKS AND	Category	L	Т	Р	Credit	Year of Introduction
386	IoT APPLICATIONS	VAC	3	1	0	4	2019

#### **Preamble:**

This course equips the learners with fundamental wireless technologies for the Internet of Things(IoT) and the IoT ecosystem. It covers the underlying concepts in wireless networks, communication mechanisms, protocols, hardware, software, and the cloud platforms for IoT. The students will be able to design smart IoT applications for real world problems..

**Prerequisite:** Sound knowledge in Data Communication, Computer Networks and Programming in C

**Course Outcomes:** After the completion of the course the students will be able to

CO1	Recognize wireless technologies required for IoT ecosystem (Cognitive Knowledge Level : Understand)
CO2	Perceive the concept of IoT and M2M architecture, IoT examples, and Data Management in IoT (Cognitive Knowledge Level : Apply)
CO3	Outline the hardware components used in IoT including Sensors, Actuators and development boards (Cognitive Knowledge Level : understand)
CO4	Explain the software components of IoT (Cognitive Knowledge Level :Understand)
CO5	Demonstrate the protocols used in IoT and build IoT Programs (Cognitive Knowledge Level : Apply)
CO6	Build IoT-based smart real-time applications such as Smart Healthcare, Smart Agriculture, Smart Environment and Smart Home (Cognitive Knowledge Level : Apply)

#### Mapping of course outcomes with program outcomes

	P01	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO1 0	PO11	PO1 2
C01												$\bigotimes$
CO2		$\bigcirc$	$\bigcirc$									$\bigcirc$

CO4       Image: Cost of the second sec	CO3	$\oslash$	$\oslash$	$\oslash$	$\oslash$	$\oslash$						$\oslash$
	CO4	$\oslash$	Ø			Ø						Ø
	CO5	$\oslash$	0	٢	٢	0				Å		Ø
	CO6	Ø	0	٢	0	0	0	K	A	A	Ŵ	

	Abstract POs Defined by National Board of Accreditation									
PO#	Broad PO	PO#	Broad PO							
PO1	Engineering Knowledge	PO7	Environment and Sustainability							
PO2	Problem Analysis	PO8	Ethics							
PO3	Design/Development of solutions	PO9	Individual and teamwork							
PO4	Conduct investigations of complex problems	PO10	Communication							
PO5	Modern tool usage	PO11	Project Management and Finance							
PO6	The Engineer and Society	PO12	Lifelong learning							

#### **Assessment Pattern**

Blooms Category	Continuous As	End Semester Examination Marks	
	Test 1 (Percentage)	Test 2 (Percentage)	
Remember	30	30	30
Understand	50	40	40
Apply	20	30	30

Analyze		
Evaluate		
Create		

**Mark Distribution** 

ABDUL KALA

Total Marks	CIE Marks	ESE Marks	ESE Duration
150	50	100	3 Hours

### **Continuous Internal Evaluation Pattern:**

Attendance		10 marks
Continuous Asse	ssment Tests	25 marks
Continuous Asse	ssment Assignment	15 marks

### **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. First Internal Examination shall be preferably conducted after completing the first half of the syllabus, and the Second Internal Examination shall be preferably conducted after completing the remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly covered module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 questions (preferably, 3 questions each from the completed modules and 1 questions (preferably, 3 questions each from the completed modules and 1 questions (preferably, 3 questions each from the completed modules and 1 questions (preferably, 3 questions each from the completed modules and 1 questions (preferably, 3 questions each from the completed modules and 1 questions (preferably, 3 questions each from the completed modules and 1 questions (preferably, 3 questions each from the completed modules and 1 questions from the partly covered module), each with 7 marks. Out of the 7 questions in Part B, a student should answer any 5.

#### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer anyone. Each question can have a maximum 2 subdivisions and carries 14 marks.

# **Syllabus**

#### ule-1 (Introduction to IoT and wireless technologies required for IoT )

Internet of Things, Role of Things and the Internet, Wireless IoT. Wireless Networks - Network Topologies, Types of Networks. Role of Wireless Standards in IoT. Protocol Stack - OSI Model, TCP/IP Model, IEEE 802 Reference Model, Protocols for Wireless IoT. Bluetooth - Transceiver, Frequency Channels, Typical Range, Access and Spread Spectrum, Modulation and Data Rate, Error Correction and Detection, Network Topology. ITU G.9959, Zwave, IEEE 802.15.4, Zigbee Specification, Thread, WiFi, 6LowPAN, IPv6, LoRaWAN.

### Module- 2 (IoT architecture, Data and Device management)

Internet of Things - IoT Architectural View, Technology Behind IoT - Server End Technology, Sources of Internet of Things, M2M Communication. IoT Application Areas. IoT Examples. IoT Data Management - Device Management Gateways. Design Principles for Web Connectivity - Web Communication Protocols for Connected Devices, Web Connectivity for Connected Devices using Gateways. Internet Connectivity Principles – Internet Connectivity, Internet based communication, IP addressing in the IoT.

### **Module- 3 (Data Acquiring and Enabling Technologies)**

Data Acquiring and Storage for IoT Sevices- Organization of Data, Big data, Acquiring Methods, Management Techniques, Analytics, Storage Technologies. Cloud Computing for Data storage - IoT Cloud based Services using Xively, Nimbits, and Other Platforms. Sensor Technologies for IoT Devices - Sensor Technology, Participatory Sensing, Industrial IoT and Automotive IoT, Actuators for Various Devices, Sensor Data Communication Protocols, Wireless Sensor network Technology

# Module-4 (Prototyping the Embedded Devices for IoT)

Embedded Computing Basics, Embedded Hardware Unit. Embedded Platforms for Prototyping -Arduino, Intel Galileo, Intel Edison, Raspberry Pi, BeagleBone, mBed. Prototyping and Designing the Software for IoT Applications- Introduction, Prototyping Embedded DeviceSoftware- Programming using Arduino, Programming for an Arduino Controlled Traffic Control Lights at a Road Junction, Basic Arduino Programs to Blink LED, Find the Distance using Ultrasonic Sensor, Estimate Room Temperature, Measuring Soil Moisture Level

#### Module 5 (Business Models and Case Studies)

Business Models and Processes using IoT. Value Creation in the Internet of Things. Cloud PaaS- Xively, Nimbits, IBM Bluemix, CISCO IoT, AWS IoT, TCS Connected AWS Platform, Case studies- Smart Home, Smart Environment, Smart healthcare, Smart agriculture

#### **Text Books**

- 1. Daniel Chew, "Wireless Internet of Things -A Guide to the lower layers", IEEE Standards and Association, IEEE Press, Wiley
- 2. Rajkamal, "Internet of Things : Architecture and Design Principles", McGraw Hill (India) Private Limited.

#### References

- 1. ArshadeepBahga, Vijay Madisetti, "Internet of Things: A hands-on approach", University Press, 2015 (First edition)
- 2. Dieter Uckelmann, Mark Harrison, Michahelles Florian (Ed.), Architecting the internet of things, Springer, 2011
- 3. Dr. Ovidiu Vermesan, Dr. Peter Friess, Internet of Things: Converging Technologies for Smart Environments and Integrated Ecosystems, River Publishers, 2013
- 4. Simon Monk, "Programming Arduino: Getting Started with Sketches", McGraw Hill Publications

# Sample Course Level Assessment Questions

#### **Course Outcome 1 (CO1):**

- 1. Compare Bluetooth and Bluetooth LE power classes
- 2. Demonstrate Zigbee Specification Protocol Stack

#### Course Outcome 2 (CO2):

- 1. What are the major components of IOT system? Briefly explain each
- 2. Correlate M2M architectural Levels with IOT architectural Levels

#### **Course Outcome 3 (CO3):**

- 1. Describe the use of GPIO pins?
- 2. What are actuators ? Mention the roles of actuators in IoT systems

#### **Course Outcome 4(CO4):**

- 1. Identify the role of HBase in Hadoop File System
- 2. Differentiate Edge computing and Distributed computing
- 3. Illustrate open protocols, tools and frameworks generally used in M2M

#### **Course Outcome 5(CO5):**

- 1. What do you mean by Arduino sketches?
- 2. Write an Arduino program to blink LED

#### **Course Outcome 6(CO6):**

- 1. How IoT technology helps TELEMEDICINE in India?
- 2. How soil moisture can be detected in Smart Agriculture?

	Model Question Paper	
QP CODE:		PAGES :2
Reg No:	TECLINIOLOCICAL	
Name:	ICUNIQUULAL	

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY SIXTH SEMESTER B.TECH DEGREE EXAMINATION(MINOR), MONTH & YEAR

# Course Code: CST 386

**Course Name: WIRELESS NETWORKS AND IOT APPLICATIONS** 

#### Max.Marks:100

**Duration: 3 Hours** 

#### **PART** A

#### Answer All Questions. Each Question Carries 3 Marks

- 1. Illustrate Role of *things* and *internet* in IoT
- 2. What is Bluetooth? Explain the range and frequency channels of Bluetooth?
- 3. List any three the features of Constrained Application Protocol (COAP).
- 4. Compare Raspberry Pi and BeagleBoard boards.
- 5. Identify the role of HBase in Hadoop File System.
- 6. Differentiate Edge computing and Distributed computing.
- 7. Give an example of Raspberry Pi applications for Industrial IoT.
- 8. What are the on-board functional units in Intel Galileo?
- 9. Interpret the concept of value creation in IoT.

10. Explain the use of PaaS in IoT Smart applications with any three examples.

(10x3=30)

			Part B	
		(A	Answer any one question from each module. Each question carries 14 Marks)	
	11.	(a)	Compare various Network topologies used in Wireless Networks.	(8)
		(b)	Describe the following wireless technologies on i) <i>Zigbee</i> ii) <i>WiFi</i> iii) <i>Thread</i> .	(6)
			OR	
	12.	(a)	Explain protocol stacks used in wireless networks for IoT applications.	(8)
		(b)	Illustrate the Architectural design of LoRaWAN.	(6)
	13.	(a)	Define M2M. Explain M2M architecture. Correlate M2M architectural levels with IoT architectural levels.	(8)
		(b)	Compare SOAP and REST protocols.	(6)
			OR	
	1	4.	(a) Estic Summarize different Online Transactions and Processing techniques.	(8)
		(b)	Identify the functions of Device-Management Gateway.	(6)
15.		(a)	2014 actuators ? Describe the roles of actuators in IoT systems.	Define (8)
		(b)	Explain the usage contexts of analog sensors and digital sensors.	(6)
			OR	
16.		(a)	Example Control of Con	How data ( <b>10</b> )
		(b)	List any four features of Xively.	(4)

17.	(a)	What do you mean by Arduino sketches?	(4)
	(b)	Write an Arduino program to blink LED	(10)
		API ABDUR KALAM	
18.	(a)	Demonstrate an example of Raspberry Pi applications for Industrial Io	T. (10)
	(b)	Compare the features of Arduino-R3 and Arduino Yun boards.	(4)
19.	(a)	Explain various tasks of a smart irrigation monitoring service.	(8)
	(b)	Demonstrate the tasks of Soil-Moisture monitoring service.	(6)
		OR	

- 20. (a) a) Mr. Kiran Mathew has been a chronic diabetic patient for the past few years. He was under regular check up at the hospital every two weeks. All of a sudden the pandemic like COVID-19 arises in the country and the government issues a lockdown for a period of two months. Illustrate how Mr. Kiran can be monitored by the health care worker using intelligent healthcare techniques.
  - (b) Mention any four sensors used in smart healthcare

#### (4)

#### **TEACHING PLAN**

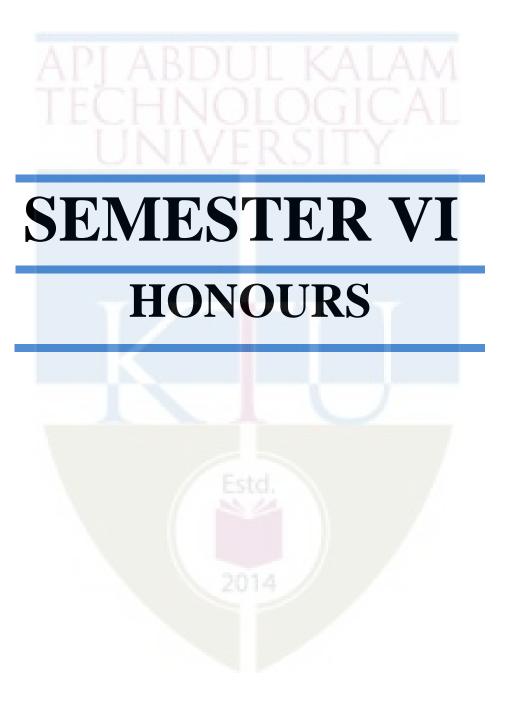
No	Contents	No of Lecture Hrs(45)				
Modu	Module – 1 ( Introduction to IoT and wireless technologies required for IoT) (8 hrs) 1, Chapter 1)					
1.1	Internet Of Things, Role of things and internet ,Wireless IoT	1				
1.2	Wireless Networks- Network Topologies-Types of Networks, Role of	1				

	Wireless standards in IoT	
1.3	Protocol Stack-OSI Model- TCP/IP Model-IEEE 802 reference model	1
1.4	Protocols for Wireless IoT-Bluetooth-Transceiver, Frequency Channels- Typical Range, Access and Spread Spectrum, Modulation and Data Rate	1
1.5	Error Correction and Detection-Network Topology.	1
1.6	ITU G.9959, Zwave, IEEE 802.15.4, Zigbee Specification	1
1.7	Thread, Wifi, 6LowPAN, IPv6	1
1.8	LoRaWAN	1
	Module- 2 (IOT architecture, Data and Device management) (9hrs)	
2.1	Internet of Things -IoT Architectural view	1
2.2	Technology Behind IOT-Server End Technology, Sources of Internet of Things	1
2.3	M2M Communication.	1
2.4	IoT Application Areas. IOT Examples.	1
2.5	IoT Data Management, Device Management Gateways.	1
2.6	Design Principles for Web Connectivity	1
2.7	Web communication protocols for connected devices,	1
2.7 2.8	Web communication protocols for connected devices,         Web connectivity for connected devices using Gateways.	1
2.8	Web connectivity for connected devices using Gateways.         Internet connectivity Principles – Internet Connectivity, Internet based	1
2.8	Web connectivity for connected devices using Gateways.         Internet connectivity Principles – Internet Connectivity, Internet based communication, IP addressing in the IoT.	1
2.8 2.9	Web connectivity for connected devices using Gateways.         Internet connectivity Principles – Internet Connectivity, Internet based communication, IP addressing in the IoT.         Module- 3 (Data Acquiring and Enabling Technologies (8 hrs)	1

	Nimbits, and other platforms.					
3.4	Cloud computing-Nimbits	1				
3.5	Sensor Technologies for IoT Devices-Sensor Technology, Participatory sensing					
3.6	Industrial IoT and Automotive IoT	1				
3.7	Actuators for various devices, Sensor data communication protocols	1				
3.8	Wireless Sensor network Technology	1				
Modu	e 4(Prototyping the Embedded Devices for IoT )(9hrs)					
4.1	Introduction, Embedded Computing Basics, Embedded Hardware Unit.	1				
4.2	Embedded Platforms for Prototyping-Arduino, Intel Galileo	1				
4.3	Intel Edison, Raspberry Pi, BeagleBone, mBed	1				
4.4	Prototyping and designing the software for IoT applications-Introduction, Prototyping embedded device software	1				
4.5	Prototyping and designing the software for IoT applications-Introduction, Prototyping embedded device software	1				
4.6	Programming concepts in Arduino	1				
4.7	Programming for an arduino controlled traffic control lights at a road junction	1				
4.8	Basic Arduino programs to blink LED, Find the distance using ultrasonic sensor	1				
4.9	Estimate room temperature, Measuring soil moisture level	1				
	Module 5 (higher level protocols and case studies)(9 hrs)					
5.1	Business Models and Processes using IOT, Value creation in the Internet of Things.	1				

5.2	Xively, Nimbits, IBM Bluemix	1
5.3	CISCO IoT, AWS IoT, TCS Connected AWS Platform	1
5.4	Case Study- Smart Environment	1
5.5	Case Study- Smart Environment	1
5.6	Case study Smart Home	1
5.7	Case study Smart Home	1
5.8	Case study Smart healthcare (Lecture I)	1
5.9	Case study Smart healthcare (Lecture II)	1
5.10	Case study -Smart agriculture (Lecture I)	1
5.11	Case study -Smart agriculture (Lecture II)	1





CST 394	NETWORK SECURITY	Category	L	Т	Р	Credits	Year of Introduction
		VAC	3	1	0	4	2019

#### **Preamble:**

The purpose of this course is to create a better understanding of the network security concepts. This course covers network security standards, email security services, web security mechanisms, firewalls and wireless security mechanisms. This course helps the learner to gain insight into the key aspects of secure network communication and enables to apply in real-life scenarios.

**Prerequisite:** A sound background in Number Theory and Cryptographic Algorithms.

**Course Outcomes:** After the completion of the course the student will be able to

CO#	Course Outcomes
CO1	Identify the key aspects of security, intrusion detection systems and digital signature schemes (Cognitive Knowledge Level: Apply)
CO2	Explain the security standards used in network communication (Cognitive Knowledge Level:Understand)
CO3	Identify the mechanisms in email security services (Cognitive Knowledge Level: Apply)
CO4	Summarize the protocols used to provide web security (Cognitive Knowledge Level: Understand)
CO5	Explain the fundamental concepts of wireless network security and firewalls (Cognitive Knowledge Level: Understand)

	PO1	PO2	PO3	PO4	PO5	PO6	<b>PO7</b>	PO8	PO9	PO10	PO11	PO12
CO1												
CO2		0		0	D		L K	A	LA	M	1	
CO3			0			9	0	61	67	XL		
<b>CO4</b>					9	9	5	TT	Y			
CO5			0									

# Mapping of course outcomes with program outcomes

	Abstract POs defined by National Board of Accreditation							
PO#	Broad PO	PO#	Broad PO					
PO1	Engineering Knowledge	PO7	Environment and Sustainability					
PO2	Problem Analysis	PO8	Ethics					
PO3	Design/Development of solutions	PO9	Individual and team work					
PO4	Conduct investigations of complex problems	PO10	Communication					
PO5	Modern tool usage	PO11	Project Management and Finance					
PO6	The Engineer and Society	PO12	Lifelong learning					

#### Assessment Pattern

Diagon's Catagony	Continuous As	sessment Tests	End Semester
Bloom's Category	Test 1 (%)	Test 2 (%)	Examination (%)
Remember	30	30	30
Understand	40	40	40
Apply	30	30	30
Analyze			
Evaluate			
Create			

Total Marks	CIE Marks	ESE Marks	ESE Duration		
150	50 100		3		
Continuous Internal Evalua	tion Pattern:	L KALA	M		
Attendance	:10	marks			
Continuous Assessment Tests	s : 25	marks			
Continuous Assessment Assig	gnment : 15	marks			

#### **Mark Distribution**

#### **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. The first series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed modules and 1 question from the should answer all questions, a student should answer any 5.

#### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 questions from each module of which a student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.

# Syllabus

### Module – 1 (Network Security Basics)

Introduction to network security - Security requirements, Challenges of security, Network security model. Malicious programs – Worms, Viruses, Trojans, Spyware, Adware. Intrusion Detection Systems (IDS) - Uses, Techniques. Digital signatures - ElGamal, Schnorr, Digital Signature Standard (DSS).

#### Module – 2 (Network Security Standards)

Kerberos v4 – Configuration, Authentication, Encryption, Message formats. Kerberos v5 – Cryptographic algorithms, Message formats. Public Key Infrastructure (PKI) – Trust models, Revocation. Real-time communication security – Perfect Forward Secrecy (PFS), Denial-of-Service protection, Endpoint identifier hiding, Live partner reassurance. Internet Protocol Security (IPSec) - Authentication Header (AH), Encapsulating Security Payload (ESP), Internet Key Exchange (IKE) phases.

#### Module – 3 (Email Security)

Introduction to email security - Security services for email, Establishing keys, Privacy, Authentication, Message integrity, Non-repudiation. Privacy Enhanced Mail (PEM) – Encryption, Source authentication and integrity protection, Message formats. Secure/Multipurpose Internet Mail Extensions (S/MIME) – Messages, Differences from PEM. Pretty Good Privacy (PGP) - Encoding, Certificate and key revocation, Anomalies, Object formats.

#### Module – 4 (Web Security)

Introduction to web security - Web security considerations, Threats. Secure Sockets Layer (SSL) – Architecture, Protocols, Transport Layer Security (TLS) – Differences from SSL. Hypertext Transfer Protocol Secure (HTTPS) – Connection initiation, Closure. Secure Shell (SSH) – Transport layer protocol, User authentication protocol, Connection protocol.

# Module – 5 (Wireless Network Security and Firewalls)

IEEE 802.11 Wireless LAN - Network components, Architectural model, Services. IEEE 802.11i wireless LAN security - Services, Phases of operation. Wired Equivalent Privacy (WEP), Wi-Fi Protected Access (WPA), WPA2, Wireless Application Protocol (WAP) – Services, Protocol architecture. Firewalls – Need for firewalls, Packet filters, Circuit-level firewalls, Application layer firewalls.

#### **Text Books**

- 1. C. Kaufman, R. Perlman and M. Speciner, "Network Security: Private Communication in a Public World", 2/e, PHI.
- 2. William Stallings, "Cryptography and Network Security Principles and Practice", 5/e, Pearson

Education Asia.

#### References

- 1. Behrouz A. Forouzan, Debdeep Mukhopadhyay, "Cryptography and Network Security", 3/e, Tata McGraw Hill.
- 2. Tyler Wrightson, "Wireless Network Security A Beginner's Guide", 2012, Tata McGraw Hill.
- 3. William Stallings, "Network Security Essentials: Applications and Standards", 4/e, Prentice Hall.
- 4. Schiller J., Mobile Communications, 2/e, Pearson Education.
- 5. Roberta Bragg et. al., "Network Security: The Complete Reference", Tata McGraw Hill.

# **Course Level Assessment Questions**

#### **Course Outcome 1 (CO1):**

- 1. Using the Schnorr digital signature scheme, let q = 83, p = 997 and d = 23. Find values for  $e_1$  and  $e_2$ .
- 2. The Digital Signature Algorithm (DSA) specifies that if the signature generation process results in a value of zero, a new value of *k* should be generated and the signature should be recalculated. Give reason.

#### Course Outcome 2 (CO2):

- 1. In Kerberos v4, the authenticator field is not of security benefit when asking the Key Distribution Center (KDC) for a ticket for Bob, but useful when logging in as Bob. Give reasons for your answer.
- 2. How does the stateless cookie protocol provide clogging protection?

#### **Course Outcome 3 (CO3):**

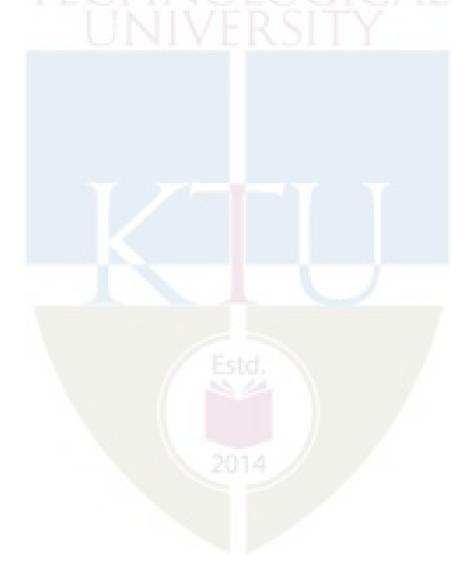
- 1. If Alice is sending an ENCRYPTED message, she first signs the message digest with her private key and then encrypts the message digest with the pre-message secret key. Why this last encryption was considered necessary for encrypted messages and not for MIC-CLEAR or MIC-ONLY?
- 2. Which security services are considered desirable in the following cases? (i) Sending a purchase order (ii) Sending a ransom note. (iii) Sending a mission description to security officials.
- 3. Explain the security mechanism used in Gmail communication.

#### **Course Outcome 4 (CO4):**

- 1. possible in SSL for the receiver to reorder SSL record blocks that arrive out of order?If so, how it can be done? If not, why?
- 2. Describe any five web security threats, their consequences and countermeasures.

# **Course Outcome 5 (CO5):**

- 1. Explain the security areas addressed by IEEE 802.11i.
- 2. Describe the advantages and disadvantages of application layer firewalls.



# **Model Question Paper**

<b>QP CODE:</b>	
Reg. No:	
Name:	

PAGES:3

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY SIXTH SEMESTER B.TECH. DEGREE (HONORS) EXAMINATION, MONTH &YEAR Course Code: CST 394

#### **Course Name: Network Security**

#### Max.Marks:100

**Duration: 3 Hours** 

#### PART A

#### Answer all Questions. Each question carries 3 Marks

- 1. Distinguish between signature-based and anomaly-based intrusion detection techniques.
- <sup>2.</sup> A trusted third party is considered as a main component in a network security model. Why?
- <sup>3.</sup> How is endpoint identifier hiding achieved in real-time communication?
- 4. Show how encryption is used to provide privacy and integrity in Kerberos v5.
- 5. End-to-end privacy is essential for e-mail security. How is this achieved?
- 6. List the four steps for preparing an EnvelopedData MIME entity.
- <sup>7.</sup> Show the operation of a Secure Sockets Layer (SSL) Record protocol.
- 8. For Secure Shell (SSH) packets, what is the advantage of not including the MAC in the scope of packet encryption?
- 9. List the three security services provided by IEEE 802.11i.
- <sup>10.</sup> Define the terms Access Point, Basic Service Set, Extended Service Set.

(10x3=30)

# Part B

#### (Answer any one question from each module. Each question carries 14 Marks)

11.	(a)	Using the ElGamal scheme, let $p = 881$ and $d = 700$ , find values for e1 and e2. Choose $r = 17$ . Find the value of S1 and S2 if $M = 400$ .	(8)
	(b)	Explain the requirements and chanenges of network security.	(6)
12.	(a)	OR In ElGamal, Schnorr and DSS, what happens if an attacker can find the value of random secret key used by the signer? Also, what happens if a user uses the same value of random secret key to sign two messages? Explain your answer for each scheme separately.	(8)
	(b)	Explain the network security model with the help of a neat diagram.	(6)
13.	(a)	Alice wishes to log into Bob's workstation remotely. List the steps involved in this communication if Kerberos v4 is used.	(7)
	(b)	How does Diffie-Hellman technique provide perfect forward secrecy using signature keys?	(7)
		OR	
14.	(a)	Explain the algorithm for Message Authentication Code (MAC) calculation and verification in Kerberos v5 rsa-md5-des.	(8)
	(b)	Compare the aggressive mode and main mode of Phase 1 Internet Key Exchange (IKE).	(6)
15.	(a)	Describe the different methods by which authentication of source is performed in email communication.	(7)
	(b)	Explain the Signed data and Clear-signed data functions provided by S/MIME.	(7)
		OR	
16.	(a)		(7)

(7) Explain the advantages of Pretty Good Privacy (PGP) over Privacy (7) Enhanced Mail (PEM).

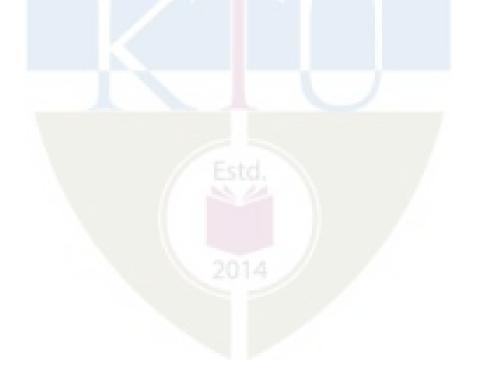
	(b)	Define non-repudiation. Describe the different ways by which it is implemented in email communication.	(7)
17.	(a)	Describe the significance of pseudo-random function of Transport Layer Security.	(7)
	(b)	Explain the four different phases of Secure Sockets Layer (SSL) HandshakeProtocol.	(7)
		OR	
18.	(a)	Describe how connection initiation and connection closure is done in Hyper Text Transfer Protocol Secure (HTTPS).	(7)
	(b)	Illustrate the sequence of events in Secure Shell (SSH) transport layer protocol packet exchanges.	(7)
19.	(a)	Explain the Discovery phase and Authentication phase of IEEE 802.11i operation.	(7)
	(b)	Why are firewalls needed? Compare the features of packet filters and circuit level firewalls.	(7)
		OR	
20.	(a)	Explain the two authentication methods used in Wired Equivalent Privacy (WEP).	(7)
	(b)	Describe the three transaction classes provided by Wireless Transaction Protocol.	(7)

# **Teaching Plan**

No	Contents								
Module - 1 (Network Security Basics) (7 hrs)									
1.1	Security requirements, Challenges of security	1							
1.2	Network security model	1							
1.3	Worms, Viruses, Trojans, Spyware, Adware	1							
1.4	Intrusion Detection Systems (IDS) uses, Techniques	1							
1.5	ElGamal digital signature	1							
1.6	Schnorr digital signature	1							
1.7	Digital Signature Standard (DSS)	1							
	Module - 2 (Network Security Standards) (12 hrs)								
2.1	Kerberos v4 configuration, Authentication	1							
2.2	Kerberos v4 encryption								
2.3	Kerberos v4 message formats								
2.4	Kerberos v5 cryptographic algorithms – rsa-md5-des, des-mac, des-mac-k								
2.5	Kerberos v5 cryptographic algorithms - rsa-md4-des, rsa-md4-des-k, Encryption for privacy and integrity	1							
2.6	Kerberos v5 message formats	1							
2.7	Public Key Infrastructure (PKI) trust models	1							
2.8	PKI revocation	1							
2.9	Perfect Forward Secrecy (PFS), Denial-of-Service protection								
2.10	Endpoint identifier hiding, Live partner reassurance								
2.11	Internet Protocol Security (IPSec) Authentication Header (AH), Encapsulating Security Payload (ESP)								

2.12	Internet Key Exchange (IKE) phases							
	Module - 3 (Email Security) (9 hrs)							
3.1	Security services for email, Establishing keys, Privacy	1						
3.2	Authentication, Message integrity, Non-repudiation	1						
3.3	Privacy Enhanced Mail (PEM) encryption, Source authentication	1						
3.4	PEM integrity protection, Message formats (Lecture 1)	1						
3.5	PEM message formats (Lecture 2)	1						
3.6	Secure/Multipurpose Internet Mail Extensions (S/MIME) – Messages, Differences from PEM	1						
3.7	Pretty Good Privacy (PGP) encoding, Certificate and key revocation, Anomalies	1						
3.8	PGP Object formats (Lecture 1)	1						
3.9	PGP Object formats (Lecture 2)	1						
	Module – 4 (Web Security)(9 hrs)							
4.1	Web security considerations, Threats, Secure Sockets Layer (SSL) architecture	1						
4.2	SSL protocols (Lecture 1)	1						
4.3	SSL protocols (Lecture 2)	1						
4.4	Transport Layer Security (TLS) differences from SSL (Lecture 1)	1						
4.5	TLS differences from SSL (Lecture 2)	1						
4.6	Hypertext Transfer Protocol Secure (HTTPS) connection initiation, Closure	1						
4.7	Secure Shell (SSH) transport layer protocol	1						
4.8	SSH user authentication protocol	1						
4.9	SSH connection protocol	1						

Module - 5 (Wireless Security and Firewalls) (8 hrs)										
5.1	IEEE 802.11 Wireless LAN network components, Architectural model, 1 Services									
5.2	IEEE 802.11i wireless LAN security services, Phases of operation (Lecture11)									
5.3	IEEE 802.11i phases of operation (Lecture 2)	1								
5.4	Wired Equivalent Privacy (WEP), Wi-Fi Protected Access (WPA), WPA2	1								
5.5	Wireless Application Protocol (WAP) services, Protocol architecture (Lecture 1)	1								
5.6	WAP protocol architecture (Lecture 2)	1								
5.7	Need for firewalls, Packet filters	1								
5.8	Circuit-level firewalls, Application layer firewalls	1								



AIT 396	MACHINE LEARNING IN COMPUTATIONAL BIOLOGY	CATEGORY	L	Т	Р	Credit	Year of Introduction
	DIOLOGI	Honors	3	1	0	4	2020

#### **Preamble:**

This course is intended to provide the learners a outlook towards application of Machine learning algorithms in the field of computational biology. This course helps the learners to apply the Machine learning methods - clustering algorithms, dimensionality reduction, decision trees, Artificial Neural Network, Support Vector Machine to the computational biology problems. Also the course discuss Challenges of Machine Learning in Computational Biology.

Prerequisite: Basic background in Bioinformatics and Machine Leaning

Course Outcomes: After the completion of the course, the student will be able to

CO 1	Describe the basic concepts of Machine Leaning, Classification, regression and clustering problems, parameters and measures ( <b>Cognitive knowledge level: Understand</b> )
CO 2	Demonstrate the clustering algorithm on computational biology problems (Cognitive knowledge level: Apply)
CO 3	Explain Dimensionality reduction techniques and Decision Trees in computational biology (Cognitive knowledge level : Apply)
CO 4	Illustrate Feature Extraction and Pattern recognition and Classification in the domain of Computational Biology analysis ( <b>Cognitive knowledge level: Apply</b> )
CO 5	Explain the role and challenges of Machine Learning in Computational (Cognitive knowledge level: Understand)

#### Mapping of course outcomes with program outcomes

	PO1	PO2	PO3	PO4	PO5	PO6	<b>PO7</b>	PO8	PO9	PO10	PO11	PO12
C01	$\oslash$	$\oslash$			×							$\oslash$
CO2	$\oslash$	Ø	Ø	$\oslash$	Ø							$\oslash$
CO3	$\oslash$	$\oslash$	$\oslash$	Ø	$\oslash$							$\oslash$
CO4	$\oslash$	$\oslash$	$\oslash$	$\oslash$								$\oslash$
CO5	$\oslash$	$\oslash$			$\oslash$							$\oslash$

PO#	Broad PO	PO#	Broad PO		
PO1	Engineering Knowledge	PO7	Environment and Sustainability		
PO2	Problem Analysis	PO8	Ethics		
PO3	Design/Development of solutions	PO9	Individual and team work		
PO4	Conduct investigations of complex problems	PO10	Communication		
PO5	Modern tool usage	PO11	Project Management and Finance		
PO6	The Engineer and Society	PO12	Life long learning		
Assessr	nent Pattern	/ER	SITY		

### **Assessment Pattern**

Bloom's Category	Continuous Asse	ssment Tests	End Semester Examination
	Test1 (%)	<b>Test2 (%)</b>	
Remember	30	30	30
Understand	50	50	50
Apply	20	20	20
Analyse	140		
Evaluate			
Create			

# **Mark Distribution**

Total	CIE	ESE Marks	ESE
Marks	Marks		Duration
150	50	100	3

# **Continuous Internal Evaluation Pattern:**

Attendance	10 marks
Continuous Assessment Tests (Average of Series Tests 1& 2)	25 marks
Continuous Assessment Assignment	15 marks

### **Internal Examination Pattern:**

Each of the two internal examinations has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 question from the partly completed module), each with 7 marks. Out of the 7 questions, a student should answer any5.

### **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contain 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 full questions from each module of which student should answer any one. Each question can have maximum 2 sub-divisions and carries 14 marks.



### Machine Learning in Computational Biology

### Module 1 (Overview of Machine Learning)

Overview of Machine Learning, fitting predictive models to data, Supervised and unsupervised learning, Classification, regression and clustering problems, Loss or cost functions. Parameters and hyperparameters, Training. validation and testing, Inductive bias and the bias variance trade-off, Use of clustering models.

### Module 2 (Clustering problems Computational Biology)

Hierarchical Clustering, Partition Clustering, Overview Model-Based Clustering, k-Means clustering algorithm, Advantages, Disadvantages, illustrative example of k-Means clustering, Clustering for creating phylogenetic trees, Using Clustering Approach to Identify Patients' Subtypes, Application of clustering algorithms on gene expression data.

### Module 3 (Supervised techniques for Computational Biology)

Proteomics Dataset, Data Pre-processing Algorithms, Dimension and Feature Subset Selection, Dimensionality reduction - Principal Component Analysis (PCA), Partial Least Square (PLS), Linear Discriminant Analysis (LDA), Protein Classification, Decision Trees in Bioinformatics, Proteomic Mass Spectra Classification Using Decision Tree Technique.

### Module 4 (Machine-Learning Algorithms for Computational Biology)

Machine-Learning Algorithms for Feature Selection from Gene Expression Data, Feature Extraction and Pattern recognition from sequence data, measures of a Feature. Artificial Neural Network (ANN) in Bioinformatics, Genetic Algorithms (GA) in Bioinformatics, Designing ANN for Bioinformatics, ANN in Protein Bioinformatics, Support Vector Machine with Feature Elimination.

### Module 5 (Scope of Machine Learning in Computational Biology)

Role of Machine Learning in Computational Biology, Creation and analysis of sequence data, Challenges of Machine Learning in Computational Biology, Data Errors, Mean Square Error Generative versus Discriminative, Approximation Versus Explanation, Single Versus Multiple Methods, Future directions of Machine Learning in Computational Biology.

### Books

1. Statistical Modelling and Machine Learning Principles for Bioinformatics Techniques, Tools, and Applications. Germany, Springer Singapore, 2020. 2. Yang, ZhengRong. Machine Learning Approaches to Bioinformatics. Singapore, World Scientific Publishing Company, 2010.

# References

- 1. Izadkhah, Habib. Deep Learning in Bioinformatics: Techniques and Applications in Practice. Netherlands, Elsevier Science, 2022.
- 2. Agapito, Giuseppe, et al. Artificial Intelligence in Bioinformatics: From Omics Analysis to Deep Learning and Network Mining. Netherlands, Elsevier Science, 2022.
- 3. Data Analytics in Bioinformatics: A Machine Learning Perspective. United States, Wiley, 2021.
- 4. Michailidis, George, et al. Introduction to Machine Learning and Bioinformatics. United Kingdom, CRC Press, 2008.
- 5. Zhang, Yanqing, and Rajapakse, Jagath C, Machine Learning in Bioinformatics, Germany, Wiley, 2009.
- 6. Baldi, Professor Pierre, et al. Bioinformatics, Second Edition: The Machine Learning Approach. India, Bradford, 2001.

# **Course Level Assessment Questions**

# **Course Outcome 1 (CO1):**

- 1. Compare and contrast Supervised and unsupervised learning
- 2. Differentiate Classification with regression with an example
- 3. Explain the parameters and hyperparameters of a model?
- 4. Summarize validation and testing in machine learning?

# Course Outcome 2 (CO2):

- 1. Write K-means algorithm and separate {5, 11, 19, 27, 23, 25, 6, 18, 2, 8, 10, 12, 31, 29, 4} into 3 clusters
- 2. Illustrate application of clustering algorithms on gene expression data
- 3. Differentiate K-means clustering and hierarchical clustering

# **Course Outcome 3 (CO3):**

- 1. Illustrate dimensionality reduction methods Principal Component Analysis (PCA), Partial Least Square (PLS), Linear Discriminant Analysis (LDA)
- 2. Explain Decision trees in Bioinformatics with a toy example.

# **Course Outcome 4 (CO4):**

- 1. Explain the process involved in feature extraction and pattern recognition from sequence data
- 2. Design and implement an ANN model for the prediction of relative solvent accessibility

# **Course Outcome 5 (CO5):**

- 1. Summarize role of Machine Learning in Computational Biology
- 2. Explain Challenges of Machine Learning approaches in Computational Biology

# Model Question Paper

**QP CODE:** 

Reg No: \_\_\_\_\_

Name: \_\_\_\_\_

PAGES: 3

### APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

### SIXTH SEMESTER B.TECH (Honors) DEGREE EXAMINATION, MONTH & YEAR

# Course Code: AIT 396

# **Course Name: MACHINE LEARNING IN COMPUTATIONAL BIOLOGY**

Max. Marks: 100

**Duration: 3 Hours** 

### PART A

	Answer All Questions. Each Question Carries 3 Marks			
1.	What does the regression line equation tell you?	(3)		
2.	2. How do you create a predictive data model using machine learning?			
3.	Write the major differences between K-means clustering and hierarchical clustering	(3)		
4.	List any three resources of Proteomics Datasets	(3)		
5.	What is the importance of using PCA before applying Machine learning method?	(3)		
6.	Draw example of an ANN architecture including 4 independent variables, one hidden layer with 3 hidden neurons and 2 dependent variables	(3)		
7.	What is the role of the Activation functions in Neural Networks?	(3)		
8.	What is Hinge Loss in SVM?	(3)		
9.	What is mean square error? how will you evaluate it?	(3)		
10.	What are discriminative machine learning models?	(10x3=30 )		
	Part B (Answer any one question from each module. Each question carries 14 Mark	s)		
11.	(a) With example, differentiate Supervised and unsupervised learning	(7)		
	(b) What is loss function and cost function in machine Learning. write the difference and example of loss function and cost function	(7)		
	OR			

12.	(a)	Define Train, Validation, and Test Datasets. how do you divide the data into Train, Validation, and Test Datasets.	(7)
	(b)	Explain Classification, regression and clustering methods with examples of each	(7)
13.	(a)	Use K Means clustering to cluster the following data into two groups. Assume cluster centroid are m1=2 and m2=4. The distance function used is Euclidean distance. { 2, 4, 10, 12, 3, 20, 30, 11, 25 }	(7)
	(b)	Illustrate with a toy example the application of clustering algorithms on gene expression data	(7)
		TFC-NOR GCAL	
14.	(a)	Explain the advantages, disadvantages of k-Means clustering	(7)
	(b)	What is the advantage of using hierarchical clustering over K means clustering? When to use the hierarchical clustering?	(7)
15.	(a)	Explain Dimension and Feature Subset Selection	(7)
	(b)	20 physicochemical properties of 100 set of proteins were given with the help of PCA, explain how will you reduce 20x100 in to Five properties (5x100) for the next level analysis	(7)
		OR	
16.	(a)	Explain how Linear Discriminant Analysis can be used for the dimensionality reduction with the help of a scenario in computational biology	(7)
	(b)	How do decision tree classifiers work? what types of problems can they solve in Computational Biology	(7)
17.	(a)	Explain the process of Feature Extraction and Pattern recognition from sequence data	(7)
	(b)	Illustrate the design of Artificial Neural Network for solving Computational Biology question	(7)
		OR	
18.	(a)	Explain crossover and mutation in genetic algorithm with an example	(7)
	(b)	Explain how to construct a support vector machine (SVM) to classify ovarian cancer from 30 individuals from the 15 features obtained from each patient.	(7)
19.	(a)	What role does machine learning and have to play in Computational Biology?	(7)
	(b)	Explain different kinds of Data Errors in Machine Learning that would happen in case of applying it in to the Computational Biology domain?	(7)
		OR	
20.	(a)	What are the advantages and disadvantages of application of machine	(7)

	learning in Computational Biology?	
(b)	"The transformation of huge volume of data into knowledge is the biggest challenge faced in computational biology" How can machine learning techniques help in this?	(7)

	TEACHING PLAN	4
No	Contents	No of Lecture (45 Hrs)
	Module 1 (Overview of Machine Learning) (9 hrs)	
1.1	Overview of Machine Learning	1
1.2	Fitting predictive models to data	1
1.3	Supervised and unsupervised learning	1
1.4	Classification, regression and clustering problems	1
1.5	Loss or cost functions	1
1.6	Proteins and peptides	1
1.7	Parameters and hyperparameters	1
1.8	Training. validation and testing	1
1.9	Inductive bias and the bias variance trade-off, Use of clustering models	1
	Module 2 (Clustering problems Computational Biology) (9 h	urs)
2.1	Hierarchical Clustering	1
2.2	Partition Clustering, Overview Model-Based Clustering	1
2.3	k-Means clustering, k-Means clustering algorithm	1
2.4	k-Means clustering advantages, disadvantages	1
2.5	illustrative example of k-Means clustering	1
2.6	Clustering for creating phylogenetic trees	1
2.7	Using Clustering Approach to Identify Patients' Subtypes	1
2.8	Application of clustering algorithms on gene expression data	1
2.9	Application of clustering algorithms on gene expression data	1
	Module 3 (Supervised techniques for Computational Biology) (	9 hrs)
3.1	Proteomics Datasets	1
3.2	Data Pre-processing Algorithms	1
3.3	Dimension and Feature Subset Selection	1

3.4	Dimensionality reduction	1
3.5	Principal Component Analysis (PCA)	1
3.6	Partial Least Square (PLS), Linear Discriminant Analysis (LDA)	1
3.7	Protein Classification case study	1
3.8	Decision Trees in Bioinformatics	1
3.9	Proteomic Mass Spectra Classification Using Decision Tree Technique	1
	TECHNOLOCICA	1

IVI (	odule 4 (Machine-Learning Algorithms for Computational Biology)	
4.1	Machine-Learning Algorithms for Feature Selection from Gene Expression Data	1
4.2	Feature Extraction and Pattern recognition from sequence data	1
4.3	Measures of a Feature	1
4.4	Artificial Neural Network (ANN) in Bioinformatics	1
4.5	Genetic Algorithms (GA) in Bioinformatics	1
4.6	Designing ANN for Bioinformatics	1
4.7	Designing ANN for Bioinformatics	1
4.8	ANN in Protein Bioinformatics	1
4.9	Support Vector Machine with Feature Elimination.	1
	Iodule 5 (Scope of Machine Learning in Computational Biology) (10	
Ν	Iodule 5 (Scope of Machine Learning in Computational Biology) (10         Role of Machine Learning in Computational Biology	) hrs)
<b>N</b> 5.1	Iodule 5 (Scope of Machine Learning in Computational Biology) (10         Role of Machine Learning in Computational Biology         Creation and analysis of sequence data	<b>) hrs</b> ) 1
<b>N</b> 5.1 5.2	Iodule 5 (Scope of Machine Learning in Computational Biology) (10         Role of Machine Learning in Computational Biology	<b>) hrs</b> ) 1 1
<b>N</b> 5.1 5.2 5.3	Iodule 5 (Scope of Machine Learning in Computational Biology) (10         Role of Machine Learning in Computational Biology         Creation and analysis of sequence data         Challenges of Machine Learning in Computational Biology	<b>) hrs</b> ) 1 1 1
<b>N</b> 5.1 5.2 5.3 5.4	Indule 5 (Scope of Machine Learning in Computational Biology) (10         Role of Machine Learning in Computational Biology         Creation and analysis of sequence data         Challenges of Machine Learning in Computational Biology         Data Errors in Machine Learning, Mean Square Error	<b>) hrs</b> ) 1 1 1 1 1
<b>N</b> 5.1 5.2 5.3 5.4 5.5	Indule 5 (Scope of Machine Learning in Computational Biology) (10         Role of Machine Learning in Computational Biology         Creation and analysis of sequence data         Challenges of Machine Learning in Computational Biology         Data Errors in Machine Learning, Mean Square Error         Generative versus Discriminative	<b>) hrs</b> ) 1 1 1 1 1 1 1
N           5.1           5.2           5.3           5.4           5.5           5.6	Indule 5 (Scope of Machine Learning in Computational Biology) (10         Role of Machine Learning in Computational Biology         Creation and analysis of sequence data         Challenges of Machine Learning in Computational Biology         Data Errors in Machine Learning, Mean Square Error         Generative versus Discriminative         Approximation Versus Explanation	<b>) hrs</b> ) 1 1 1 1 1 1 1 1

AIT 398	IMAGE AND VIDEO PROCESSING	Category	L	Т	Ρ	Credit	Year of Introduction
		HONORS	3	1	0	4	2020

### Preamble:

This course enables the learners to understand how digital images are stored and processed. The learners are exposed to different spatial and frequency domain methods for image enhancement, image restoration techniques, morphological operations that could be performed on digital images and also various image and video compression techniques. The course also gives an introduction to the basics of video processing and video segmentation.

Prerequisite: Advanced Computer Graphics, Advanced Concepts in Computer Vision

### Mapping of course outcomes with program outcomes

CO1	Summarize the steps of digital image processing and pixel relationships. (Cognitive Knowledge Level: Understand)
CO2	Apply spatial and frequency domain methods for image enhancement. (Cognitive Knowledge Level: Apply
CO3	Apply restoration techniques and morphological operations on digital images. (Cognitive Knowledge Level: Apply)
CO4	Compare different methods for digital image and video compression. (Cognitive Knowledge Level: Apply)
CO5	Understand the basics of video processing and video segmentation. (Cognitive Knowledge Level: Understand)

### Mapping of course outcomes with program outcomes

	PO 1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO1 0	P011	PO1 2
CO1												

CO2	$\oslash$	$\oslash$	$\oslash$	$\oslash$	Ø	$\oslash$					Ø
CO3	$\oslash$	$\oslash$	$\oslash$	$\oslash$	$\oslash$	$\oslash$					$\oslash$
CO4		0	Ø	0	0	0	D	Γ. A. 1	Å	. A	$\oslash$
CO5	$\oslash$		1	ΥĎ	2	R	-		2	VI I	Ø
		1 L	ŭ	N	N	FF	S	T	Y	1L	

	Abstract POs defined by National Board of Accreditation									
PO#		Broad PO	PO#	Broad PO						
PO1	Engi	neering Knowledge	PO7	Environment and Sustainability						
PO2	Prob	lem Analysis	PO8	Ethics						
PO3	Desi	gn/Development of solutions	PO9	Individual and team work						
PO4		duct investigations of plex problems	PO10	Communication						
PO5	Mod	ern tool usage	P011	Project Management and Finance						
PO6	The	Engineer and Society	PO12	Life long learning						

### **Assessment Pattern**

Bloom's	Continuou	s Assessment Tests	End Semester Examination
Categor y	Test 1 (%)	Test 2 (%)	Marks (%)
Remember	30	30	30
Understand	30	30	30
Apply	40	40	40
Analyze			
Evaluate			
Create			

Total Marks	CIE Marks	ESE Marks	ESE Duration						
150 A	DI A 50 DI	100	3 hrs						
Continuous Internal Evaluation Pattern:									
Attendance		10 r	narks						

# **Mark Distribution**

# Continuous Assessment Tests(Average of Internal Tests 1 & 2)25 marksContinuous Assessment Assignment15 marks

# **Internal Examination Pattern**

Each of the two internal examinations has to be conducted out of 50 marks. First series test shall be preferably conducted after completing the first half of the syllabus and the second series test shall be preferably conducted after completing the remaining part of the syllabus. There will be two parts: Part A and Part B. Part A contains 5 questions (preferably, 2 questions each from the completed modules and 1 question from the partly completed module), having 3 marks for each question adding up to 15 marks for part A. Students should answer all questions from Part A. Part B contains 7 questions (preferably, 3 questions each from the completed modules and 1 questions should answer all questions from the partly completed module), each with 7 marks. Out of the 7 questions, a student should answer any 5.

# **End Semester Examination Pattern:**

There will be two parts; Part A and Part B. Part A contains 10 questions with 2 questions from each module, having 3 marks for each question. Students should answer all questions. Part B contains 2 full questions from each module of which student should answer any one. Each question can have a maximum of 2 subdivisions and carries 14 marks.

Syllabus

Module - 1

**Fundamentals of Image processing**: Basic steps of Image processing system, sampling and quantization of an Image, basic relationship between pixels and connectivity.

**Image Enhancement**: Spatial Domain methods - Gray level transformations, Histogram Processing, Fundamentals of Spatial Filtering, Smoothing Spatial filters, Sharpening Spatial filters.

### Module -2

**Image Transforms**: Unitary transforms, 2D Discrete Fourier Transform, Discrete Cosine Transform (DCT), Discrete Wavelet transforms.

**Frequency Domain methods**: Basics of filtering in frequency domain, image smoothing, image sharpening, homomorphic filtering.

### Module - 3

**Image Restoration:** Image degradation/Restoration model, Noise models, Restoration in presence of noise only - spatial filtering, Periodic Noise reduction by frequency domain filtering.

**Morphological Operations**: Erosion, Dilation, Opening, Closing, Hit-or-miss transformation, Boundary extraction.

### Module - 4

**Image compression fundamentals** – Coding Redundancy, spatial and temporal redundancy.

**Compression models** : Lossy and Lossless, Huffman coding, Arithmetic coding, LZW coding, run length coding, Bit Plane coding, JPEG standards.

### Module - 5

Video processing: Basics of Video Processing: Analog video, Digital Video.

Video segmentation: Introduction to video segmentation, Change detection.

**Video Compression**: Introduction to video compression, video compression based on motion compensation, Search for motion vectors, H.261 standard, Transform coding, predictive coding-MPEG.

### **Text Books**

1. Gonzalez and Woods, "Digital Image Processing", 3rd edition, Pearson, 2009.

2. Li, Ze-Nian, Mark S. Drew, and Jiangchuan Liu. "Fundamentals of multimedia", Pearson Prentice Hall, 2004.

3. Bovik, Alan C. "Handbook of image and video processing", Academic press, 2010.

### **Reference Books**

1. David A. Forsyth & Jean Ponce, Computer vision – A Modern Approach, Prentice Hall, 2002.

2. Richard Szeliski, Computer Vision: Algorithms and Applications, Springer.

3. Maheshkumar H Kolekar, "Intelligent Video Surveillance Systems: An Algorithmic Approach", CRC Press.

4. Francesco Camastra, Alessandro Vinciarelli, "Machine Learning for Audio, Image and Video Analysis: Theory and Applications", Springer 2015.

5. M. Tekalp ,"Digital video Processing", Prentice Hall International

6. Relf, Christopher G., "Image acquisition and processing with LabVIEW", CRC press

7 Chris Solomon, Toby Breckon ,"Fundamentals of Digital Image Processing A Practical Approach with Examples in Matlab", John Wiley & Sons,

8. Yao wang, Joem Ostarmann and Ya – quin Zhang, "Video processing and communication ",1st edition, PHI

# **Course Level Assessment Questions**

### Course Outcome1 (CO1):

- 1. Illustrate how the image is digitized by sampling and quantization.
- 2. Let  $V = \{1,2\}$  and compute the length of the shortest 4-, 8-, and m path between p and q. If a particular path does not exist between these two points explain why.

1	2	1q
2	0	2
2	1	1
0	1	2
	2	2 0 2 1

### Course Outcome 2(CO2):

р

1. Determine whether the given matrix is unitary or not:

 $\mathbf{A} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}.$ 

2. Explain any five properties of 2D Fourier Transform.

### Course Outcome 3(CO3):

- 1. Discuss how restoration is done in digital images.
- 2. Explain with examples the different morphological operations applied to images.

### Course Outcome 4(CO4): .

- 1. With suitable examples, clearly bring out the need for compression in images and videos.
- 2. Discuss any one method for finding motion vectors.

### **Course Outcome 5(CO5):**

- 1. Explain any one technique used for segmenting a video.
- 2. Compare and contrast analog video and digital video in multimedia.

# **Model Question Paper**

QP C	ODE:
------	------

Reg No: \_\_\_\_\_

Name: \_\_\_\_\_

PAGES:3

### APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY

### SIXTH SEMESTER B.TECH DEGREE EXAMINATION, MONTH & YEAR

Course Code: AIT 398

**Course Name: Image and Video Processing** 

### Max. Marks : 100

### **Duration: 3 Hours**

# PART A

### Answer All Questions. Each Question Carries 3 Marks

- 1. Explain bit plane slicing and contrast stretching.
- 2.
  - Discuss about pixel relationships.

3. Find the 4 order forward and inverse DFT for the following image segment:



- 4. Define DCT. Write the properties of DCT.
- 5. Discuss hit or miss transformation with appropriate examples.
- 6. Explain about the morphological operation dilation.
- 7. Explain the significance of image compression.
- 8. Distinguish between lossy and lossless compression.
- 9. Discuss the significance of change detection.
- 10. Explain how transform coding is used in compression algorithms.

(10x3=30 )

### Part B

(Answer any one question from each module. Each question carries 14 Marks)

11. (a) Perform histogram specification of the following 3 bit gray scale image whose gray level distribution is given as follows.

(9)

# Input image

Gray level	0	1	2	3	4	5	6	7
No. of Pixels	8	10	10	2	12	16	4	2
Farget im	age	H	N	JL		JIC	A	
Gray Level	0	JI	2	3	4	5	6	7
No. of Pixels	0	0	0	0	20	20	16	8

(b) Design Laplacian filter for image enhancement in spatial domain.

OR

12.	(a)	What is histogram equalization? Explain the procedure for histogram equalization.	(7)
	(b)	Explain the gray level transformation functions: a) image negatives and b) log transformation c) power law transformation.	(7)
13.	(a)	Compute the 2D DFT of the 4 X 4 grayscale image given below.	(4)
	(b)	Explain about smoothing and sharpening frequency domain filters.	(10)
		OR	

14. (a)	Explain Butterworth filters for image smoothening and image sharpening.	(4)
(b)	Explain the steps followed in frequency domain filtering?	(5)

(5)

**15.** (a) Apply opening and closing operation on the image sample A given below (10) with structuring element B

A= ( 1	0	0	0	0	and $B = \Box 1 1 1 \Box$
0	1	0	0	0	
0	0	1	0	0	
0	0	0	1	0	
Lo	0	0	0	1)	

(b) Explain Morphological operations a) opening b) closing with suitable (4) examples.

### OR

16.	(a)	Discuss about different noise models.	(7)
	(b)	Explain how periodic noise reduction can be done using frequency domain filtering.	(7)
17.	(a)	Comment on JPEG compression standard.	(8)
	(b)	Discuss on run-length encoding with the help of an example.	(6)
		OR Estd.	
18.	(a)	Explain LZW coding with the help of a suitable example.	(8)
	(b)	Illustrate the concept of arithmetic coding.	(6)
19.	(a)	Compare and contrast MPEG video coding and H.261 standard.	(7)
	(b)	Explain video segmentation with an example.	(7)
		OR	

20. (a) Illustrate how motion compensation is used in video compression. (7)

(b) With the help of a neat block diagram explain predictive coding methods. (7)

	APJ AB Teaching Plan ALAM	
No	Contents Contents	No. of Lecture Hours (44 hrs)
	Module – 1 (7 hours)	
1.1	Fundamentals of Image processing: Basic steps of Image processing system, Sampling and quantization of an Image.	1 hour
1.2	Basic relationship between pixels and connectivity.	1 hour
1.3	Image Enhancement: Gray level transformations	1 hour
1.4	Histogram, Histogram Equalization	1 hour
1.5	Histogram specification	1 hour
1.6	Fundamentals of Spatial Filtering	1 hour
1.7	Smoothing Spatial filters	1 hour
1.8	Sharpening Spatial filters	1 hour
	Module-2 (8 hours)	
2.1	Image Transforms: Unitary transforms.	1 hour
2.2	2D Discrete Fourier Transform	1 hour
2.3	Discrete Cosine Transform (DCT)	1 hour

2.4	Discrete Wavelet transforms	1 hour
2.5	Basics of filtering in frequency domain	1 hour
2.6	Image smoothing	1 hour
2.7	Image sharpening	1 hour
2.8	Homomorphic filtering.	1 hour
	Module-3 (9 hours)	
3.1	Image Restoration: Image degradation/Restoration model	1 hour
3.2	Noise models	1 hour
3.3	Restoration basics	1 hour
3.4	Restoration in presence of noise only - spatial filtering	1 hour
3.5	Periodic Noise reduction by frequency domain filtering.	1 hour
3.6	Morphological Operations: basics	1 hour
3.7	Erosion, Dilation, Opening, Closing	1 hour
3.8	Hit-or-miss transformation Est d.	1 hour
3.9	Boundary extraction.	1 hour
	Module-4 (10 hours)	
4.1	Image compression fundamentals - Coding Redundancy	1 hour
4.2	Spatial and temporal redundancy.	1 hour
4.3	Compression models : Lossy and Lossless	1 hour

4.4	Huffman coding	1 hour
4.6	Arithmetic coding	1 hour
4.7	LZW coding	1 hour
4.8	Run length coding	1 hour
4.9	Bit Plane coding,	1 hour
4.10	JPEG standards	1 hour
	Module-5 (10 hours)	
5.1	Basics of Video Processing: Analog video, Digital Video.	1 hour
5.2	Video segmentation: Introduction to video segmentation	1 hour
5.3	Change detection.	1 hour
5.4	Introduction to video compression	1 hour
5.5	Video compression based on motion compensation	1 hour
5.6	Search for motion vectors	1 hour
5.7	Transform coding	1 hour
5.8	Predictive coding	1 hour
5.9	MPEG standards	1 hour
5.10	H.261 standard	1 hour