



OFFICE OF THE DEAN ACADEMICS

## SCHEME & SYLLABUS OF UNDERGRADUATE DEGREE COURSE

## B. TECH. ARTIFICIAL INTELLIGENCE & MACHINE LEARNING

III YEAR (V & VI Semester)



Effective for the students admitted in year 2021-22 and onwards Approved by ...... academic council meeting held on ......





## OFFICE OF THE DEAN ACADEMICS

# Teaching & Examination Scheme B. Tech. (Artificial Intelligence & Machine Learning) 3rd Year – V Semester

(Effective for the students admitted in year 2021-22 and onwards)

| S. No. | Category | Course<br>Code | Course Title                               |    | Hour |     | Exam<br>Hours |     | Mark |       | Credit |
|--------|----------|----------------|--|----|------|-----|---------------|-----|------|-------|--------|
|        |          | Code           |  | L  | Т    | P   | Hours         | IA  | ETE  | Total |        |
|        |          |                |  |    |      |     |               |     |      |       |        |
|        |          |                | TH   | EO | KY   |     |               |     |      |       |        |
| 1      |          | 5AM4-01        | Operating Systems                          | 3  | -    | -   | 3             | 30  | 70   | 100   | 3      |
| 2      |          | 5AM4-02        | Computer Organization and Architecture     | 3  | -    | -   | 3             | 30  | 70   | 100   | 3      |
| 3      | DC       | 5AM4-03        | Computer Networks                          | 3  | -    | -   | 3             | 30  | 70   | 100   | 3      |
| 4      |          | 5AM4-04        | Machine Learning and its Applications      | 3  | -    | -   | 3             | 30  | 70   | 100   | 3      |
| 5      |          | 5AM4-05        | Mathematical Foundation<br>Course          | 3  | -    | -   | 3             | 30  | 70   | 100   | 3      |
| 6      |          | 5AM5-11        | Information Security Systems               | 2  | -    | -   | 3             | 30  | 70   | 100   | 2      |
|        | DE       | 5AM5-12        | Introduction to Data Science               |    |      |     |               |     |      |       |        |
|        |          | 5AM5-13        | Distributed Systems                        |    |      |     |               |     |      |       |        |
| 7      |          | 5AM5-14        | Cloud Computing                            | 2  | -    | -   | 3             | 30  | 70   | 100   | 2      |
|        |          | 5AM5-15        | Introduction to Blockchain                 |    |      |     |               |     |      |       |        |
|        |          | 5AM5-16        | Data Mining and Warehousing                |    |      |     |               |     |      |       |        |
|        |          | Sub To         | otal                                       | 19 | 00   | 00  | -             | 210 | 490  | 700   | 19     |
|        |          |                | PRACTICAL &                                | SE | SSI  | ON. | AL            |     |      |       |        |
| 8      |          | 5AM4-21        | Machine Learning and<br>Neural Network Lab | -  | -    | 2   | _             | 60  | 40   | 100   | 1      |
| 9      | DC       | 5AM4-22        | R Programming Lab                          | -  | -    | 2   | -             | 60  | 40   | 100   | 1      |
| 10     | 1        | 5AM4-23        | Data Visualization Lab                     | -  | -    | 2   | -             | 60  | 40   | 100   | 1      |
| 11     | UI       | 5AM7-30        | Industrial Training                        | -  | -    | 1   | -             | 60  | 40   | 100   | 3      |
| 12     | CCA      | 5AM8-00        | SODECA / Co-Curricular<br>Activity         | -  | -    | -   | -             | -   | 100  | 100   | 1      |
|        | •        | Sub To         | otal                                       | 00 | 00   | 07  | -             | 240 | 260  | 500   | 7      |
|        |          | Tota           | ıl   | 19 | 00   | 07  | -             | 450 | 750  | 1200  | 26     |

L = Lecture, T = Tutorial, P = Practical, IA = Internal Assessment, ETE = End Term Exam, Cr = Credits





## OFFICE OF THE DEAN ACADEMICS

# Teaching & Examination Scheme B. Tech. (Artificial Intelligence & Machine Learning) 3<sup>rd</sup> Year – VI Semester

(Effective for the students admitted in year 2021-22 and onwards)

| S. No. | Category     | Course<br>Code | Course Title                                    |           | lour |     | Exam<br>Hours |            | Mark |       | Credit |
|--------|--------------|----------------|---|-----------|------|-----|---------------|------------|------|-------|--------|
|        |              |                |   | L         | T    | P   |               | IA         | ЕТЕ  | Total |        |
|        |              |                | TH  | EOI       | RY   |     |               |            |      |       |        |
| 1      |              | 6AM4-01        | Compiler Design                                 | 3         | -    | _   | 3             | 30         | 70   | 100   | 3      |
| 2      | -            | 6AM4-02        | Design and Analysis of<br>Algorithms            | 3         | -    | -   | 3             | 30         | 70   | 100   | 3      |
| 3      | DC           | 6AM4-03        | Big Data Analytics                              | 3         | -    | -   | 3             | 30         | 70   | 100   | 3      |
| 4      |              | 6AM4-04        | Digital Image Processing                        | 3         | -    | -   | 3             | 30         | 70   | 100   | 3      |
| 5      | <del> </del> | 6AM4-05        | Deep Learning                                   | 3         | -    | -   | 3             | 30         | 70   | 100   | 3      |
| 6      |              | 6AM5-11        | Internet of Things                              | 2         | -    | -   | 3             | 30         | 70   | 100   | 2      |
|        | DE           | 6AM5-12        | Pattern Recognition                             |           |      |     |               |            |      |       |        |
|        |              | 6AM5-13        | GPU Computing                                   |           |      |     |               |            |      |       |        |
|        |              | Sub To         | otal  | <b>17</b> | 00   | 00  |               | <b>180</b> | 420  | 600   | 17     |
|        |              |                | PRACTICAL                                       | &         | SES  | SIC | NAL           |            |      |       |        |
| 7      |              | 6AM4-21        | Design and Analysis of<br>Algorithms Lab        | -         | -    | 2   | -             | 60         | 40   | 100   | 1      |
| 8      | DC           | 6AM4-22        | Big Data Analytics Lab                          | _         | -    | 2   | -             | 60         | 40   | 100   | 1      |
| 9      |              | 6AM4-23        | Digital Image Processing<br>Lab                 | -         | -    | 2   | -             | 60         | 40   | 100   | 1      |
| 10     | UI           | 6AM7-50        | Innovation and Design Thinking Hands-on Project | -         | -    | 3   | -             | 60         | 40   | 100   | 2      |
| 11     | CCA          | 6AM8-00        | SODECA / Co-Curricular<br>Activity              | _         | -    | _   | -             | -          | 100  | 100   | 2      |
|        |              | Sub To         | otal  | 00        | 00   | 09  | -             | 240        | 260  | 500   | 7      |
|        |              | Tota           | 1   | 17        | 00   | 09  | -             | 420        | 680  | 1100  | 24     |

L = Lecture, T = Tutorial, P = Practical, IA = Internal Assessment, ETE = End Term Exam, Cr = Credits





## OFFICE OF THE DEAN ACADEMICS

| V Semester                 |                                     |  |  |  |  |
|----------------------------|-------------------------------------|--|--|--|--|
| B. Tech. (Artificia        | al Intelligence & Machine Learning) |  |  |  |  |
| 5AM4-01: Operating Systems |                                     |  |  |  |  |
| Credit: 3                  | Max. Marks: 100 ( IA:30, ETE:70)    |  |  |  |  |
| 3L+0T+ 0P                  | End Term Exams: 3 Hours             |  |  |  |  |

Course Objectives: As a result of successfully completing this course, students will:

- Learn about how Operating System is Important for Computer System.
- Learn about different types of Operating Systems and their services.
- Learn different process scheduling algorithms and synchronization techniques to achieve better performance of a computer system.
- Learn about device and device management.
- Learn about the concept of memory management and virtual memory.
- Learn about the concept of file system.

Course Outcomes: Upon successful completion of the course the students will be able to

- **CO-1**: Analyze basic concepts of operating systems and their structures.
- **CO-2**: Analyze various issues related to inter-process communication like process synchronization and critical section.
- **CO-3**: Synthesize the concepts of I/O management, file system implementation, scheduling, resource management and deadlocks.
- **CO-4**: Interpret the issues and challenges of memory management.
- **CO-5**: Understand protection and security issues related to the operating system.

| S. No. | Contents   | Hours |
|--------|--|-------|
| 1      | Introduction to OS and Process Management:   | 9     |
|        | Introduction to operating systems, operating system structure, system calls, Process concept,      |       |
|        | Operations on processes, cooperating processes, inter process communication, mutual                |       |
|        | exclusion, critical section problem, Synchronization hardware, wait and signal procedures,         |       |
|        | Semaphores, Classic problems of synchronization, critical regions, Monitors, process               |       |
|        | scheduling and algorithms, threads, multithreading.  |       |
|        | <b>CPU Scheduling</b> : Scheduling criteria, Scheduling algorithms, Multiple processor scheduling, |       |
|        | Real time scheduling   |       |
| 2      | Memory Management:   | 8     |
|        | Background, Swapping, Contiguous memory allocation, Paging, Segmentation, Segmentation             |       |
|        | with paging. Virtual Memory, Demand paging, Page replacement policies, Allocation of               |       |
|        | frames, Thrashing, case study.   |       |
| 3      | Deadlock and Device Management:  | 9     |
|        | Deadlock: System model, Deadlock characterization, Methods for handling deadlocks,                 |       |
|        | Deadlock prevention, Deadlock avoidance, Deadlock detection, Recovery from deadlock.               |       |
|        | Device management: devices and their characteristics, device drivers, device handling, disk        |       |
|        | scheduling algorithms, Swap space management.  |       |
| 4      | File Systems and Its Implementation:   | 7     |
|        | File System Interface, File concepts, Access methods, Directory structure, File system             |       |
|        | mounting, Directory implementation, Allocation methods, Free space management -                    |       |
|        | efficiency and performance, recovery, log structured file systems                                  |       |
| 5      | Protection and Case Studies:   | 7     |
|        | <b>Protection</b> : Goals of protection, Principles of protection, Domain of protection, Access    |       |
|        | matrix, Implementation of access matrix, Access control, Revocation of access rights, file         |       |
|        | security, user authentication  |       |





## OFFICE OF THE DEAN ACADEMICS

| Total  | 40 |
|--|----|
| Inter-process communication, Case studies of Real Time and Mobile OS.                |    |
| Process management; Scheduling; Memory management; File systems, Input and output;   |    |
| Case Study: Linux Operating System Linux history; Design principles; Kernel modules; |    |

- 1. Silberschatz, Galvin, and Gagne, "Operating System Concepts", Wiley India Pvt Ltd.
- 2. Modern Operating Systems, Andrew S. Tanenbaum, Herbert Bos, Pearson Education India; Fourth edition 2016. ISBN-13:978-9332575776
- 3. Operating Systems: Internals and Design Principles William Stallings, Pearson Education India; 7 edition (2013). ISBN-13: 978-9332518803
- 4. Gary Nutt, "Operating Systems", Third Edition, Pearson Education
- 5. Operating Systems: A Design-Oriented Approach, Charles Crowley, International edition, McGraw-Hill Education (ISE Editions). ISBN-13 978 0071144629





## OFFICE OF THE DEAN ACADEMICS

| V Semester B. Tech. (Artificial Intelligence & Machine Learning) |                                  |  |  |  |  |
|--|----------------------------------|--|--|--|--|
| D. Tech. (Artificial   | intenigence & Machine Learning)  |  |  |  |  |
| 5AM4-02: Computer Organization and Architecture                  |                                  |  |  |  |  |
| Credit: 3  | Max. Marks: 100 ( IA:30, ETE:70) |  |  |  |  |
| 3L+0T+ 0P  | End Term Exams: 3 Hours          |  |  |  |  |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- Learn the principles of computer organization and basic architectural concepts.
- Understand the basics of instructions sets and their impact on processor design.
- Demonstrate an understanding of the design of the functional units of a digital computer system.
- Evaluate cost performance and design trade-offs in designing and constructing a computer processor including memory.
- Design a pipeline for consistent execution of instructions with minimum hazards.
- Recognize and manipulate representations of numbers stored in digital computers.

#### **Course Outcomes:**

Upon successful completion of the course, students will be able to

- **CO-1:** Study of the basic structure and operation of a digital computer system.
- **CO-2:** Analysis of the design of arithmetic & logic unit and understanding of the fixed point and floating point arithmetic operations.
- **CO-3:** Implementation of control unit techniques and the concept of Pipelining.
- **CO-4**: Understanding the hierarchical memory system, cache memories and virtual memory.
- **CO-5**: Understanding the different ways of communicating with I/O devices and standard I/O interfaces.

| S. No. | Contents  | Hours |
|--------|---|-------|
| 1      | Introduction: Objective, scope and outcome of the course.                               | 1     |
| 2      | Register Transfer and Micro-operations: Register Transfer Language (RTL), Bus and       | 9     |
|        | Memory Transfers, Arithmetic Micro-Operations, Logic Micro-Operations, Shift Micro-     |       |
|        | Operations, Arithmetic Logic Shift Unit (ALU).  |       |
| 3      | Basic Computer Organization and Design: Instruction Codes, Computer Registers,          | 8     |
|        | Computer Instructions, Timing and Control, Instruction Cycle, Register-Reference and    |       |
|        | Memory- Reference Instructions, Input-Output and Interrupt, Design of Basic Computer.   |       |
| 4      | Central Processing Unit: General Register Organization, Stack Organization, Instruction | 8     |
|        | Format, Addressing Modes, Data Transfer and Manipulation, Program Control, Reduced      |       |
|        | Instruction Set Computer (RISC) and Complex Instruction Set Computer (CISC).            |       |
| 5      | Pipeline and Vector Processing: Flynn's Taxonomy, Parallel Processing, Pipelining,      | 8     |
|        | Arithmetic Pipeline, Instruction Pipeline.  |       |
|        | Computer Arithmetic: Signed Magnitude Binary Numbers - Addition and Subtraction,        |       |
|        | Multiplication- Booth Multiplication Algorithm, Array Multiplier, Division Algorithm.   |       |
| 6      | Input-Output Organization: Input-output Interface Modes of Transfer, Daisy Chaining     | 8     |
|        | Priority, Direct Memory Access (DMA), Input-Output Processor (IOP)- CPU-IOP             |       |
|        | Communication.  |       |
|        | Memory Organization: Memory Hierarchy, Main Memory, Auxiliary Memory, Associative       |       |
|        | Memory, Cache Memory, Virtual Memory.   |       |
|        | Total   | 42    |





## OFFICE OF THE DEAN ACADEMICS

- 1. M. Morris Mano, Computer System Architecture, Pearson
- 2. Carl Hamacher, Zvonko Vranesic, Safwat Zaky Computer Organization, McGraw-Hill, Fifth Edition, Reprint 2012
- 3. John P. Hayes, Computer Architecture and Organization, Tata McGraw Hill, Third Edition, 1998. Reference books
- 4. William Stallings, Computer Organization and Architecture-Designing for Performance, Pearson Education, Seventh edition, 2006.
- 5. Behrooz Parahami, "Computer Architecture", Oxford University Press, Eighth Impression, 2011.
- 6. David A. Patterson and John L. Hennessy, "Computer Architecture-A Quantitative Approach", Elsevier, a division of reed India Private Limited, Fifth edition, 2012
- 7. Structured Computer Organization, Tannenbaum(PHI)





## OFFICE OF THE DEAN ACADEMICS

| V Semester B. Tech. (Artificial Intelligence & Machine Learning) |                                   |  |  |  |  |
|--|-----------------------------------|--|--|--|--|
| 5AM4-03: Computer Networks                                       |                                   |  |  |  |  |
| Credit: 3  | Max. Marks: 100 ( IA:30, ETE:70 ) |  |  |  |  |
| 3L+0T+ 0P  | End Term Exams: 3 Hours           |  |  |  |  |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- Become familiar with layered communication architectures (OSI and TCP/IP models).
- Understand different services offered by various OSI and TCP/IP model layers.
- Understand the client/server model and key application layer protocols.
- Understand the concept of unreliable data transfer and its role in communication.
- Understand the concepts of reliable data transfer and how TCP implements these concepts.
- Know the principles of congestion control and trade-offs in fairness and efficiency.
- Understand the role and concept of routing in communication.
- Understand the basics of error detection, including parity, checksums, and CRC.
- Familiarize the student with current topics such as security, network management, sensor networks, and/or other topics.

### **Course Outcomes:**

Upon successful completion of the course, students will be able to

- **CO-1:** Understand basic computer network technology.
- **CO-2:** Understand OSI and TCP/IP reference model and working of each layer of these reference models.
- **CO-3:** Obtain the skills of subnetting and routing mechanisms.
- **CO-4:** Address design and implementation aspects of various essential network protocols and its integration into network-based applications.

| S. No. | Contents  | Hours |
|--------|---|-------|
| 1      | <b>Introduction:</b> history and development of computer networks, networks topologies. Layering and protocols. OSI and TCP/IP Protocol Stacks, Basics of packet, circuit and virtual circuit switching.  | 6     |
|        | <b>Physical Layer:</b> Guided Transmission media: twisted pairs, coaxial cable, fiber optics, Wireless transmission.  |       |
| 2      | <b>Data link layer:</b> Design issues, framing, Error detection and correction. Elementary data link protocols: simplex protocol, A simplex stop and wait protocol for an error-free channel, A simplex stop and wait protocol for noisy channel. Sliding Window protocols: A one-bit sliding window protocol, A protocol using Go-Back-N, A protocol using Selective Repeat, Example data link protocols. Medium Access sub layer: The channel allocation problem, Multiple access protocols: ALOHA, Carrier sense multiple access protocols, collision free protocols. Wireless LANs, Data link layer switching, Ethernet bridging. | 8     |
| 3      | <b>Network Layer:</b> Design issues, Routing algorithms, shortest path routing, Flooding, Hierarchical routing, Broadcast, Multicast, distance vector routing, link state routing, Congestion Control Algorithms, Quality of Service, Internetworking, Fragmentation, The Network layer in the internet, IP addressing, IPv4, IPv6. CIDR, NAT, Basics of IP support protocols (ARP, DHCP, ICMP)   | 8     |
| 4      | <b>Transport Layer:</b> Transport Services, Elements of Transport protocols, Connection management, Error and Flow Control, Congestion Control, TCP and UDP protocols, Sockets.   | 7     |
| 5      | Application Layer: Domain name system, Electronic Mail; the World Wide Web, HTTP,   | 7     |





## OFFICE OF THE DEAN ACADEMICS

|    | FTP, Streaming audio and video.   |    |
|----|---|----|
| 6. | Current Topics Related to Computer Network: Basic overview of the role and working of     | 6  |
|    | topic such as Software-defined Networks, Wireless Sensor Networks and Internet of Things, |    |
|    | Cyber-physical systems  |    |
|    | Total   | 42 |

- 1. Computer Networks, Andrew S. Tanenbaum and David J Wetherall, 5th Edition. Pearson publication.
- 2. Computer Networking: A Top-Down Approach Featuring the Internet, James F Kurose and Keith W Ross. Pearson publication.
- 3. Computer Networking: A Top-Down Approach, Behrouz A. Forouzan, Firouz Mosharraf, TMH.
- 4. Data Communications and Networking Behrouz A. Forouzan. 4th Edition TMH.
- 5. Computer Networks: A Systems Approach, 5th Ed., LL Peterson, BS Davie, Morgan-Kauffman, 2011.
- 6. Cryptography and Network Security, Principles and Practice, 5th Ed., W Stallings, Prentice-Hall, 2010
- 7. Internet of Things: A Hands-on Approach , by Arshdeep Bagha and Vijay Madisetti, Universities Press, 2015, ISBN: 9788173719547
- 8. Fundamentals of Cyber-Physical Systems https://eprints.whiterose.ac.uk/173235/1/Chapter%201 .%20 Fundamentals%20of%20 Cyber-Physical %20Systems.pdf
- 9. Cyber-Physical Systems and Internet of Things https://nvlpubs.nist.gov/nistpubs/SpecialPublications/NIST.SP.1900-202.pdf





## OFFICE OF THE DEAN ACADEMICS

| V Semester<br>B. Tech. (Artificial Intelligence & Machine Learning) |                                  |  |  |  |  |  |
|---|----------------------------------|--|--|--|--|--|
| 5AM4-04: Machine Learning and its Applications                      |                                  |  |  |  |  |  |
| Credit: 3   | Max. Marks: 100 ( IA:30, ETE:70) |  |  |  |  |  |
| 3L+0T+ 0P   | End Term Exams: 3 Hours          |  |  |  |  |  |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- Develop a comprehensive understanding of machine learning and AI concepts and principles.
- Acquire skills to apply machine learning techniques to real-world problems.
- Ability to design and develop AI models for complex problem-solving.
- Cultivate critical thinking and problem-solving abilities in the context of machine learning and AI.

### **Course Outcomes:**

Upon successful completion of the course, students will be able to

- **CO-1:** Analyze methods and theories in the field of machine learning and provide an introduction to the basic principles, techniques, and applications of machine learning, classification tasks, decision tree learning.
- **CO-2:** Apply decision tree learning, Bayesian learning and artificial neural network in real world problems.
- **CO-3:** Understand the use of genetic algorithms and genetic programming.
- **CO-4:** Apply inductive and analytical learning with related domain theories.
- **CO-5**: Compare different learning models and algorithms and utilize existing machine learning algorithms to design new algorithms.

| S. No. | Contents   | Hours |
|--------|--|-------|
| 1      | Introduction: Objective, scope and outcome of the course.                                      | 1     |
| 2      | Artificial Neural Network: Neural network representation, Neural Networks as a                 | 9     |
|        | paradigm for parallel processing, Linear discrimination, Pairwise separation, Gradient         |       |
|        | Descent, Logistic discrimination, Perceptron, Training a perceptron, Multilayer                |       |
|        | perceptron, Back propagation Algorithm. Recurrent Networks, Dynamically modifying              |       |
|        | network structure.   |       |
| 3      | <b>Decision Tree Learning:</b> Decision tree representation, appropriate problems for decision | 8     |
|        | tree learning, Univariate Trees (Classification and Regression), Multivariate Trees, Basic     |       |
|        | Decision Tree Learning algorithms, Hypothesis space search in decision tree learning,          |       |
|        | Inductive bias in decision tree learning, Issues in decision tree learning.                    |       |
| 4      | Genetic Algorithms: Basic concepts, Hypothesis space search, Genetic programming,              | 6     |
|        | Models of evolution and learning, Parallelizing Genetic Algorithms.                            |       |
| 5      | Bayesian Learning: Bayes theorem and concept learning, Bayes optimal classifier,               | 8     |
|        | Gibbs algorithms, Naive Bayes Classifier, Bayesian belief networks, The EM algorithm.          |       |
|        | Design of Machine Learning: Guidelines for machine learning experiments, Factors,              |       |
|        | Response, and Strategy of experimentation  |       |
| 6      | Analysis of Machine Learning: Cross-Validation and Resampling methods, measuring               | 8     |
|        | classifier performance, Hypothesis testing, Assessing a classification algorithm's             |       |
|        | performance, Comparing two classification algorithms, Comparing multiple algorithms:           |       |
|        | Analysis of variance, Comparison over multiple datasets.                                       |       |
|        | Total  | 40    |





## OFFICE OF THE DEAN ACADEMICS

- 1. Mitchell M., T., Machine Learning, McGraw Hill (1997) 1stEdition.
- 2. Kevin Murphy, Machine Learning: A Probabilistic Perspective, MIT Press, 2012
- 3. Ethem Alpaydin, Introduction to Machine Learning, MIT Press, 2004
- 4. Tom Mitchell, Machine Learning, McGraw Hill, 1997.
- 5. Pat Langley, Elements of Machine Learning, Morgan Kaufmann Publishers, Inc. 1995. ISBN 1-55860-301-8
- 6. Bishop M., C., Pattern Recognition and Machine Learning, Springer-Verlag (2011) 2ndEdition.
- 7. Michie D., Spiegelhalter J. D., Taylor C. C., Campbell, J., Machine Learning, Neural and Statistical Classification. Overseas Press (1994).





## OFFICE OF THE DEAN ACADEMICS

| V Semester<br>B. Tech. (Artificial Intelligence & Machine Learning) |                         |
|---|-------------------------|
| 5AM4-05: Mathematical Foundation Course                             |                         |
| Credit: 3 Max. Marks: 100 ( IA:30, ETE:70)                          |                         |
| 3L+0T+ 0P   | End Term Exams: 3 Hours |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- Able to learn and understand the fundamental concepts in probability & statistics, Liner methods, Basic of vector space and Linear Transformations.
- Able to perform test of hypothesis
- Learn about Mathematics foundation of various ML, AI and DS methods.

#### **Course Outcomes:**

Upon successful completion of the course, students will be able to

CO-1:. Able to Understand sampling theory and sampling distributions

**CO-2:** Able to Understand multivariate statistics

**CO-3:** To make aware of the Sampling and Test of Hypothesis.

CO-4: Able to Understand about basic linear algebra

**CO-5:** Able to Understand the Linear Transformations and its use in AI.

| S. No. | Contents  | Hours |
|--------|---|-------|
| 2      | Sampling Theory: Population and Sample, Statistical inference, Sampling with and without replacement, Random samples, Population parameters, Sample statics, Sampling distributions, Sample mean, Sampling distribution of means, Sample variances, Sampling distribution of variances, Case where population variances is unknown, Unbiased estimates and efficient estimates, the point estimate and Interval Estimates, & Confidence Interval, sampling distributions, Confidence Interval estimates of population parameters, Confidence intervals for the variance of a Normal distribution, Maximum likelihood estimates.  Introduction to Multivariate Statistics-Degree of Relationship among Variables-Review of | 10    |
|        | Univariate and Bivariate Statistics-Screening Data Prior to Analysis-Missing Data, Outliers, Normality, Linearity, and Homoscedasticity   |       |
| 3      | <b>Test of Hypothesis and Significance</b> Statistical hypothesis, Null and Alternate hypothesis, the test of hypothesis and significance, Type I and Type II errors, Level of Significance, Tests involving the Normal distribution, Goodness of fit, Test of Independence, Permutations and Randomization Test, t-test/z-test (one sample, independent, paired), One-Tailed and Two-Tailed tests, P-value. Special tests of significance for large samples and small samples (F, chi-square, z,), Analysis of Variance and Covariance (ANOVA & ANCOVA) - Multivariate Analysis of Variance and Covariance (MANOVA & MANCOVA)  | 10    |
| 4      | <b>Basics of Linear Algebra:</b> System of Linear Equations, Vector space and subspaces (definition, examples, and concepts of basis), Linear mappings, Matrices, Eigenvalues and Eigenvectors Norms, Inner Product, Orthogonally, Spectral Decomposition, Singular value Decomposition, Low-rank Approximation, Projection, Principal Component Analysis and Generative Model  | 8     |
| 5      | <b>Linear Transformations:</b> Linear Transformations and Matrices for Linear Transformation, Kernel and Range of a Linear Transformations, Change of Basis <b>Information Theory:</b> Entropy, cross-entropy, KL divergence, mutual information  | 6     |
|        | Total   | 40    |





## OFFICE OF THE DEAN ACADEMICS

University Press (1st edition) 2020

- 2. S. Axler, Linear Algebra Done Right. Springer International Publishing (3rd edition) 2015
- 3. E. Kreyszig, Advanced Engineering Mathematics, John Wiley and Sons, Inc., U.K. (10th Edition) 2015
- 4. R. A. Johnson, I. Miller, and J. E. Freund, "Miller & Freund's Probability and Statistics for Engineers", Prentice Hall PTR, (8th edition) 2011
- 5. E. Walpole, R. H. Mayers, S. L. Mayers, and K. Ye, (2007), Probability and Statistics for Engineers and Scientists,8th Edition, Pearson Education
- 6. Douglas C. Montgomery, (2012), Applied Statistics and Probability for Engineers, 5th Edition, Wiley India,
- 7. Spiegel, M. R., Schiller, J., and Srinivasan, R. A., (2010), Probability & Statistics, 3rdEdition, Tata McGraw Hill,
- 8. S.C. Gupta & V.K. Kapoor, "Fundamentals of Mathematical Statistics", Sultan Chand & Sons
- 9. Sheldon M. Ross, "Introduction to Probability and Statistics for Engineers and Scientists", Academic Press.





## OFFICE OF THE DEAN ACADEMICS

| V Semester<br>B. Tech. (Artificial Intelligence & Machine Learning) |                         |
|---|-------------------------|
| 5AM5-11: Information Security Systems                               |                         |
| Credit: 2 Max. Marks: 100 ( IA:30, ETE:70                           |                         |
| 2L+0T+ 0P   | End Term Exams: 3 Hours |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- Understand vulnerability in a computer system.
- Understand basic concept of how to protect and design private network.
- Understand how to protect security of information.
- Use theoretical and practical knowledge in securing data transfer and authentication.

#### **Course Outcomes:**

Upon successful completion of the course, students will be able to

- CO-1: Identify the security issues in the network and resolve it
- **CO-2:** Analyze the vulnerabilities in any computing system and hence be able to design a security solution.
- **CO-3:** Evaluate security mechanisms using rigorous approaches by key ciphers and Hash functions.
- **CO-4:** To understand various network security applications, IPSec, Firewall, IDS, Web Security, Email Security and Malicious software

| S. No. | Contents  | Hours |
|--------|---|-------|
| 1      | Introduction: Objective, scope and outcome of the course.                                 |       |
| 2      | Security Attacks, Security Services, Security Mechanisms and Principles, Security goals,  | 5     |
|        | Malicious software, Worms, Viruses, Trojans, Spyware, Botnets                             |       |
| 3      | Basic of Cryptography: Symmetric and asymmetric cryptography, cryptographic hash          | 5     |
|        | functions, authentication and key establishment, Message Authentication Codes (MACs),     |       |
|        | digital signatures.   |       |
| 4      | Security Vulnerabilities: DoS attacks, Buffer Overflow, Race Conditions, Access           | 5     |
|        | Control Problems, Spoofing and Sniffing attacks.  |       |
| 5      | Internet Security: TCP/IP Security, Secure Sockets Layer (SSL), Transport Layer           | 5     |
|        | Security (TLS), HTTPS, Secure Shell (SSH), IPsec, Email Security, DNS Security,           |       |
|        | Authentication Protocols  |       |
| 6      | Web Security: Phishing attack, SQL Injection, Securing databases and database access,     | 7     |
|        | Cross Site Scripting Attacks, Cookies, Session Hijacking, E-commerce security             |       |
|        | System Security: Firewalls, Types: Packet filter (stateless, stateful), Application layer |       |
|        | proxies, Firewall Location and Configurations, Intruders, Intrusion Detection System.     |       |
|        | Total   | 28    |

- 1. Stallings, W., Network Security Essentials, Prentice Hall (2017) 6th Edition.
- 2. Cheswick, R., W., Bellovin, M., S., and Rubin, D., A., Firewalls and Internet Security, Addison-Wesley Professional (2003) 2nd Edition.
- 3. Graves, K., Certified Ethical Hacking Study Guide, Sybex (2010) 1st Edition.
- 4. Stallings, W., Cryptography and Network Security, Prentice Hall (2013), 6th Edition.





## OFFICE OF THE DEAN ACADEMICS

| V Semester<br>B. Tech. (Artificial Intelligence & Machine Learning) |                         |
|---|-------------------------|
| 5AM5-12: Introduction to Data Science                               |                         |
| Credit: 2 Max. Marks: 100 ( IA:30, ETE:70                           |                         |
| 2L+0T+ 0P   | End Term Exams: 3 Hours |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- To understand EDA, inference and regression techniques.
- Apply Matrix decomposition techniques to perform data analysis.
- Understand concepts and importance of data pre-processing techniques.
- Importance and application of Machine Learning Algorithms.
- Knowledge of acquiring data through web-scraping and data APIs

#### **Course Outcomes:**

Upon successful completion of the course, students will be able to

**CO-1:** Utilize EDA, inference and regression techniques.

**CO-2:** Utilize Matrix decomposition techniques to perform data analysis.

CO-3: Apply data pre-processing techniques.

**CO-4:** Apply Basic Machine Learning Algorithms.

**CO-5:** Acquire data through web-scraping and data APIs.

| S. No. | Contents  | Hours |
|--------|---|-------|
| 1      | Introduction: Objective, scope and outcome of the course.   | 1     |
| 2      | Introduction to data analysis: Introduction and importance of data science. Big Data Analytics, Business intelligence vs Big data, Current landscape of analytics, Exploratory Data Analysis (EDA), statistical measures, Basic tools (plots, graphs and summary statistics) of EDA, Data Analytics Lifecycle, Discovery, Data Visualization Principles of Data Visualization   | 6     |
| 3      | Introductory hypothesis testing and statistical inference: Introduction to Hypothesis Testing, Central Limit Theorem, A/B testing. Identifying Potential Data Sources  Linear regression - Introduction to simple linear regression, multiple linear regression, least squares principle, exploratory vs. inferential viewpoints, Model generalizability, cross validation, and using categorical variables in regression, logistic regression, Multiple correlation, Partial correlation | 5     |
| 4      | Linear Algebra Basics- Matrices to represent relations between data, Linear algebraic operations on matrices – Matrix decomposition: Singular Value Decomposition (SVD) and Principal Component Analysis (PCA).   | 5     |
| 5      | Data Pre-processing and Feature Selection - Data cleaning - Data integration - Data Reduction - Data Transformation and Data Discretization, Feature Generation and Feature Selection, Feature Selection algorithms: Filters- Wrappers - Decision Trees - Random Forests  | 6     |
| 6      | Basic Machine Learning Algorithms - Classifiers - Decision tree - Naive Bayes - k-Nearest Neighbors (k-NN), k-means - SVM Association Rule mining - Ensemble methods  | 5     |
|        | Total   | 28    |

- 1. Mining of Massive Datasets. v2.1, Jure Leskovek, Anand Rajaraman and Jefrey Ullman., Cambridge University Press. (2019)
- 2. Doing Data Science, Straight Talk From The Frontline, Cathy O'Neil and Rachel Schutt, O'Reilly
- 3. Python for Data Analysis: Data Wrangling with Pandas, NumPy, and IPython Wes McKinney, O'Reilly





## OFFICE OF THE DEAN ACADEMICS

| _ | _ |   |   |   |
|---|---|---|---|---|
| n | М | ~ | 1 | • |
|   |   |   |   |   |

4. Hands-On Machine Learning with Scikit-Learn and TensorFlow: Concepts, Tools, and Techniques to Build Intelligent Systems, Aurélien Géron, O'Reilly Media





## OFFICE OF THE DEAN ACADEMICS

| V Semester B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
|--|-------------------------|--|
| 5AM5-13: Distributed Systems                                     |                         |  |
| Credit: 2 Max. Marks: 100 (IA:30, ETE:70 )                       |                         |  |
| 2L+0T+ 0P  | End Term Exams: 3 Hours |  |

**Course Objectives**: As a result of successfully completing this course, students will:

- To Understand hardware and software issues in modern distributed systems.
- To get knowledge in distributed architecture, naming, synchronization, consistency and replication, fault tolerance, security, and distributed file systems.
- To analyze the current popular distributed systems such as peer-to-peer (P2P) systems will also be analyzed.

Course Outcomes: Upon successful completion of the course, students will be able to

- **CO-1:** To understand the foundations of distributed systems.
- **CO-2:** To learn issues related to clock Synchronization and the need for global state in distributed systems.
- **CO-3:** To learn distributed mutual exclusion and deadlock detection algorithms.
- **CO-4:** To understand the significance of agreement, fault tolerance and recovery protocols in Distributed Systems.

CO-5: To learn the characteristics of peer-to-peer and distributed shared memory systems

| S. No. | Contents   | Hours |
|--------|--|-------|
| 1      | Introduction: Objective, scope and outcome of the course.  | 1     |
| 2      | <b>Distributed Systems:</b> Features of distributed systems, nodes of a distributed system, Distributed computation paradigms, Model of distributed systems, Types of Operating systems: Centralized Operating System, Network Operating Systems, Distributed Operating Systems and Cooperative Autonomous Systems, design issues in distributed operating systems. Systems Concepts and Architectures: Goals, Transparency, Services, Architecture Models, Distributed Computing Environment (DCE). | 5     |
| 3      | Theoretical issues in distributed systems: Notions of time and state, states and events in a distributed system, time, clocks and event precedence, recording the state of distributed systems.  Concurrent Processes and Programming: Processes and Threads, Graph Models for Process Representation, Client/Server Model, Time Services, Language Mechanisms for Synchronization.  | 5     |
| 4      | <b>Distributed Process Scheduling:</b> A System Performance Model, Static Process Scheduling with Communication, Dynamic Load Sharing and Balancing, Distributed Process Implementation.  Distributed File Systems: Transparencies and Characteristics of DFS, DFS Design and implementation, Transaction Service and Concurrency Control  | 5     |
| 5      | <b>Distributed Shared Memory</b> : Non-Uniform Memory Access Architectures, Memory Consistency Models, Multiprocessor Cache Systems, Distributed Shared Memory, Implementation of DSM systems.   | 6     |
| 6      | <b>Distributed Agreement:</b> Concept of Faults, failure and recovery, Replicated Data Management: concepts and issues, Database Techniques, Atomic Multicast, and Update Propagation. CORBA case study: Introduction, Architecture, CORBA RMI, CORBA Services.  | 6     |
|        | Total  | 28    |





## OFFICE OF THE DEAN ACADEMICS

- 1. Distributed Systems, Principles and Paradigms, 2nd edition by Andrew S. Tanenbaum and Maarteen Van Steen, Pearson Education, (ISBN-13: 978- 0132392273), 2013 IT-89
- 2. Distributed System: Concepts and Design, 5th edition by Coulouris, Dollimore, Kindberg, Pearson Ed, (ISBN-13: 978-0132143011), 2013
- 3. Distributed Algorithms: Principles, Algorithms, and Systems by A. D. Kshemkalyani and M. Singhal, (ISBN-13: 978-0521189842), 2013





## OFFICE OF THE DEAN ACADEMICS

| V Semester<br>B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
|---|-------------------------|--|
| 5AM5-14: Cloud Computing  |                         |  |
| Credit: 2 Max. Marks: 100 ( IA:30, ETE:70                           |                         |  |
| 2L+0T+ 0P   | End Term Exams: 3 Hours |  |

**Course Objectives**: As a result of successfully completing this course, students will:

- The fundamental ideas behind Cloud Computing, the evolution of the paradigm, its applicability; benefits
- The basic ideas and principles in data center design; cloud management techniques and cloud software deployment considerations;
- Different CPU, memory and I/O virtualization techniques in cloud

Course Outcomes: Upon successful completion of the course, students will be able to

- **CO-1:** Explain the core concepts of the cloud computing paradigm
- **CO-2:** Discuss system, network and storage virtualization and outline their role in enabling the cloud computing system model.

CO-3: Understanding security architecture of cloud infrastructure

| S. No. | Contents  | Hours |
|--------|---|-------|
| 1      | Introduction: Objective, scope and outcome of the course.   | 1     |
| 2      | Cloud Computing: Nutshell of cloud computing, Enabling Technology, Historical development, Vision, feature Characteristics and components of Cloud Computing. Challenges, Risks and Approaches of Migration into Cloud. Ethical Issue in Cloud Computing, Evaluating the Cloud's Business Impact and economics, Future of the cloud. Networking Support for Cloud Computing.        | 5     |
| 3      | - · · ·   |       |
| 4      | <b>Virtualization Technology:</b> Definition, Understanding and Benefits of Virtualization. Implementation Level of Virtualization, Virtualization Structure/Tools and Mechanisms, Hypervisor VMware, KVM, Xen. Virtualization of CPU, Memory, I/O Devices, Virtual Cluster and Resources Management, Virtualization of Server, Desktop, Network, and Virtualization of data-centre | 5     |
| 5      |   |       |
| 6      | <b>Data Security in Cloud:</b> Business Continuity and Disaster Recovery, Risk Mitigation, Understanding and Identification of Threats in Cloud, SLA-Service Level Agreements, Trust Management   | 6     |
|        | Total   | 28    |

- 1. Rajkumar Buyya, James Broberg, Andrzej M. Goscinski: "Cloud Computing: Principles and Paradigms", Wiley, 2011
- Rajkumar Buyya, Christian Vecchiola, S Thamarai Selvi, Mastering Cloud Computing, Tata McGraw Hill, 2013
- 3. Barrie Sosinsky: "Cloud Computing Bible", Wiley-India, 2010





## OFFICE OF THE DEAN ACADEMICS

- 4. Ronald L. Krutz, Russell Dean Vines: "Cloud Security: A Comprehensive Guide to Secure Cloud Computing", Wiley-India, 2010
- 5. Tim Mather, Subra Kumara swamy, Shahed Latif, Cloud Security and Privacy: An Enterprise Perspective on Risks and Compliance, O'Reilly Media, 2009.





## OFFICE OF THE DEAN ACADEMICS

| V Semester B. Tech. (Artificial Intelligence & Machine Learning) |  |
|--|--|
| 5AM5-15: Introduction to Blockchain                              |  |
| Credit: 2 Max. Marks: 100 ( IA:30, ETE:70 )                      |  |
| 2L+0T+ 0P End Term Exams: 3 H                                    |  |

### **Course Objectives**: As a result of successfully completing this course, students will:

- The students should be able to understand a broad overview of the essential concepts of blockchain technology.
- To familiarize students with Bitcoin protocol followed by the Ethereum protocol to lay the foundation necessary for developing applications and programming.
- Students should be able to learn about different types of blockchain and consensus algorithms.

Course Outcomes: Upon successful completion of the course, students will be able to

- **CO-1:** To explain the basic notion of distributed systems.
- **CO-2:** To use the working of an immutable distributed ledger and trust model that defines blockchain.
- **CO-3:** To illustrate the essential components of a blockchain platform.

| S. No. | Contents  | Hours |
|--------|---|-------|
| 1      | Introduction: Objective, scope and outcome of the course.   | 1     |
| 2      | Basics: The Double-Spend Problem, Byzantine Generals' Computing Problems, Public-Key Cryptography, Hashing, Distributed Systems, Distributed Consensus.   | 5     |
| 3      | Technology Stack: Blockchain, Protocol, Currency. Bitcoin Blockchain: Structure, Operations, Features, Consensus Model, Incentive Model   | 5     |
| 4      | Ethereum Blockchain: Smart Contracts, Ethereum Structure, Operations, Consensus Model, Incentive Model.   | 5     |
| 5      | Tiers of Blockchain Technology: Blockchain 1.0, Blockchain 2.0, Blockchain 3.0, Types of Blockchain: Public Blockchain, Private Blockchain, Semi-Private Blockchain, Sidechains.  | 6     |
| 6      | Types of Consensus Algorithms: Proof of Stake, Proof of Work, Delegated Proof of Stake, Proof Elapsed Time, Deposite-Based Consensus, Proof of Importance, Federated Consensus or Federated Byzantine Consensus, Practical Byzantine Fault Tolerance. Blockchain Use Case: Supply Chain Management. | 6     |
| Total  |   | 28    |

- 1. Kirankalyan Kulkarni, Essentials of Bitcoin and Blockchain, Packt Publishing.
- 2. Anshul Kaushik, Block Chain & Crypto Currencies, Khanna Publishing House.
- 3. Tiana Laurence, Blockchain for Dummies, 2nd Edition 2019, John Wiley & Sons.
- 4. Mastering Blockchain: Deeper insights into decentralization, cryptography, Bitcoin, and popular Blockchain frameworks by Imran Bashir, Packt Publishing (2017).
- 5. Blockchain: Blueprint for a New Economy by Melanie Swan, Shroff Publisher O'Reilly Publisher Media; 1st edition (2015).





## OFFICE OF THE DEAN ACADEMICS

| V Semester B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
|--|-------------------------|--|
| 5AM5-16: Data Mining and Warehousing                             |                         |  |
| Credit: 2 Max. Marks: 100 ( IA:30, ETE:70 )                      |                         |  |
| 2L+0T+ 0P  | End Term Exams: 3 Hours |  |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- To introduce the fundamental processes data warehousing and major issues in data mining
- To impart the knowledge on various data mining concepts and techniques that can be applied to text mining, web mining etc.
- To develop the knowledge for application of data mining and social impacts of data mining.

#### **Course Outcomes:**

Upon successful completion of the course, students will be able to

- **CO-1:** Interpret the contribution of data warehousing and data mining to the decision-support systems.
- **CO-2:** Prepare the data needed for data mining using preprocessing techniques.
- **CO-3:** Extract useful information from the labeled data using various classifiers.
- **CO-4:** Compile unlabeled data into clusters applying various clustering algorithms.
- CO-5: Discover interesting patterns from large amounts of data using Association Rule Mining

| S. No. | Contents  | Hours |
|--------|---|-------|
| 1      | Introduction: Objective, scope and outcome of the course.   | 1     |
| 2      | <b>Introduction to Data Mining:</b> Introduction to data mining-Data mining functionalities-<br>Steps in data mining process- Classification of data mining systems, Major issues in data<br>mining. Data Wrangling and Preprocessing: Data Preprocessing: An overview-Data<br>cleaning-Data transformation and Data discretization                                 | 5     |
| 3      | <b>Predictive Modeling:</b> General approach to classification-Decision tree induction- Bayes classification methods- advanced classification methods: Bayesian belief networks Classification by Backpropagation- Support Vector Machines-Lazy learners  | 6     |
| 4      | <b>Descriptive Modeling:</b> Types of data in cluster analysis-Partitioning methods- Hierarchical methods-Advanced cluster analysis: Probabilistic model-based clustering- Clustering high dimensional data-Outlier analysis  | 5     |
| 5      | <b>Discovering Patterns and Rules:</b> Frequent Pattern Mining: Basic Concepts and a Road Map - Efficient and scalable frequent item set mining methods: Apriori algorithm, FP-Growth algorithm- Mining frequent item sets using vertical data format- Mining closed and max patterns Advanced Pattern Mining: Pattern Mining in Multilevel, Multidimensional Space | 5     |
| 6      | Data Mining Trends and Research Frontiers: Other methodologies of data mining: Web mining Temporal mining-Spatial mining-Statistical data mining- Visual and audio data mining- Data mining applications- Data mining and society: Ubiquitous and invisible data mining- Privacy, Security, and Social Impacts of data mining                                       | 6     |
|        | Total   | 28    |

- 1. Jiawei Han and Micheline Kamber, Data Mining: Concepts and Techniques, Morgan Kaufmann Publishers, third edition ,2013
- 2. Pang-Ning Tan, Michael Steinbach, Anuj Karpatne, Vipin Kumar, Introduction to Data Mining, second edition, Pearson, 2019
- 3. Ian. H. Witten, Eibe Frank and Mark. A. Hall, Data Mining: Practical Machine Learning Tools and Techniques, third edition, 2017





## OFFICE OF THE DEAN ACADEMICS

- 4. Alex Berson and Stephen J. Smith, Data Warehousing, Data Mining & OLAP, Tata McGraw Hill Edition, Tenth Reprint, 2008.
- 5. Hand, D., Mannila, H. and Smyth, P. Principles of Data Mining, MIT Press: Massachusetts third edition, Pearson, 2013





## OFFICE OF THE DEAN ACADEMICS

| V Semester  |                         |  |
|---|-------------------------|--|
| B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
| 5AM4-21: Machine Learning and Neural Network Lab      |                         |  |
| Credit: 1 Max. Marks: 100 ( IA:60, ETE:40             |                         |  |
| 0L+0T+ 2P   | End Term Exams: 2 Hours |  |

Course Objectives: As a result of successfully completing this course, students will:

- Gain hands-on experience in implementing and applying machine learning algorithms and techniques.
- Develop skills in preprocessing and analyzing data for machine learning tasks.
- Acquire proficiency in using popular machine learning frameworks and libraries.
- Learn to evaluate and optimize machine learning models through practical experimentation.

Course Outcomes: Upon successful completion of the course, students will be able to

- **CO-1:** Develop practical skills in implementing and training machine learning models using various algorithms and techniques.
- **CO-2:** Gain hands-on experience in preprocessing and analyzing real-world datasets for machine learning tasks.
- **CO-3:** Acquire proficiency in using industry-standard tools and libraries for machine learning and AI development.
- **CO-4:** Learn to evaluate model performance, interpret results, and make data-driven decisions.
- **CO-5:** Apply ethical considerations and address potential biases in the design and implementation of machine learning systems.

| S. No. | List of Experiments   |  |
|--------|---|--|
| 1      | Implement and demonstrate the FIND-Salgorithm for finding the most specific hypothesis based on a       |  |
|        | given set of training data samples. Read the training data from a .CSV file.                            |  |
| 2      | For a given set of training data examples stored in a .CSV file, implement and demonstrate the          |  |
|        | Candidate-Elimination algorithm to output a description of the set of all hypotheses consistent with    |  |
|        | the training examples.  |  |
| 3      | Write a 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   |  |
| 4      | Build an Artificial Neural Network by implementing the Backpropagation algorithm and test the same      |  |
|        | using appropriate data sets   |  |
| 6      | Write a program to implement the naïve Bayesian classifier for a sample training data set stored as a   |  |
|        | .CSV file. Compute the accuracy of the classifier, considering few test data sets.                      |  |
| 7      | Write a program to construct a Bayesian network considering medical data. Use this model to             |  |
|        | demonstrate the diagnosis of heart patients using standard Heart Disease Data Set. You can use          |  |
|        | Java/Python ML library classes/API.   |  |
| 8      | Apply EM algorithm to cluster a set of data stored in a .CSV file. Use the same data set for clustering |  |
|        | using k-Means algorithm. Compare the results of these two algorithms and comment on the quality of      |  |
|        | clustering. You can add Java/Python ML library classes/API in the program.                              |  |
| 9      | Write a program to implement k-Nearest Neighbour algorithm to classify the iris data set. Print both    |  |
|        | correct and wrong predictions. Java/Python ML library classes can be used for this problem.             |  |

- 1. Mitchell M., T., Machine Learning, McGraw Hill (1997) 1stEdition.
- 2. Kevin Murphy, Machine Learning: A Probabilistic Perspective, MIT Press, 2012





## OFFICE OF THE DEAN ACADEMICS

- 3. Ethem Alpaydin, Introduction to Machine Learning, MIT Press, 2004
- 4. Tom Mitchell, Machine Learning, McGraw Hill, 1997.
- Pat Langley , Elements of Machine Learning, Morgan Kaufmann Publishers, Inc. 1995. ISBN 1-55860-301-8
- 6. Bishop M., C., Pattern Recognition and Machine Learning, Springer-Verlag (2011) 2ndEdition.
- 7. Michie D., Spiegelhalter J. D., Taylor C. C., Campbell, J., Machine Learning, Neural and Statistical Classification. Overseas Press (1994).





## OFFICE OF THE DEAN ACADEMICS

| V Semester B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
|--|-------------------------|--|
| 5AM4-22: R Programming Lab                                       |                         |  |
| Credit: 1 Max. Marks: 100 ( IA:60, ETE:40 )                      |                         |  |
| 0L+0T+ 2P  | End Term Exams: 2 Hours |  |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- Explain critical R programming concepts
- Demonstrate how to install and configure RStudio and Apply OOP concepts in R programming
- Explain the use of data structure and loop functions
- Analyze data and generate reports based on the datasets

#### **Course Outcomes:**

Upon successful completion of the course, students will be able to

- **CO-1:** Show the installation of R Programming Environment.
- **CO-2:** Utilize and R Data types for developing programs.
- **CO-3:** Make use of different R Data Structures.
- **CO-4:** Develop programming logic using R Packages.
- CO-5: Analyze the datasets using R programming capabilities.

| 20 2. That y 20 the datasets using it programming capabilities. |   |  |
|---|---|--|
| S. No.  | List of Experiments   |  |
| 1   | Point-to-point transformation. This laboratory experiment provides for thresholding an image and the evaluation of its histogram. Histogram equalization. This experiment illustrates the relationship among the intensities (gray levels) of an image and its histogram. |  |
| 2   | Geometric transformations. This experiment shows image rotation, scaling, and translation. Two-dimensional Fourier transform  |  |
| 3   | Linear filtering using convolution. Highly selective filters.   |  |
| 4   | Ideal filters in the frequency domain. Non Linear filtering using convolutional masks. Edge detection. This experiment enables students to understand the concept of edge detectors and their operation in noisy images.  |  |

- 1. R Programming for Data Science, Roger D Peng, Lean Publication, 2016
- 2. R for Data Science: Import, Tidy, Transform, Visualize, and Model Data by Hadley Wickham, O'RELLY, 2017
- 3. Hands-On Programming with R: Write Your Own Functions and Simulations, Garrett Goleman, O'RELLY, 2014 <a href="http://cran.r-project.org(link is external">http://cran.r-project.org(link is external)</a>





## OFFICE OF THE DEAN ACADEMICS

| V Semester B. Tech. (Artificial Intelligence & Machine Learning)   |                                   |  |
|--|-----------------------------------|--|
| 5AM4-23: Data Visualization Lab  |                                   |  |
| Credit: 1  | Max. Marks: 100 ( IA:60, ETE:40 ) |  |
| 0L+0T+ 2P  | End Term Exams: 2 Hours           |  |
| Course Objectives: As a result of successfully completing this course, students will:  Handle data and data visualisations to demonstrate an understanding of ethical considerations |                                   |  |

### **Course Outcomes:**

Upon successful completion of the course, students will be able to

surrounding data (including data storage, citation, and protection).

- CO-1: To introduce students to the fundamental problems, concepts, and approaches in the design and analysis of data visualization systems.
- **CO-2**: Analyze data using exploratory visualization
- CO-3: Build commonly requested types of visualizations as well as more advanced visualizations using ground-up customization.
- CO-4: Create useful, performing visualizations from real-world data sources, including large and complex datasets

| S. No. | List of Experiments  |
|--------|--|
| 1      | Learn how to import data from various sources such as SQL database, CSV, XML, XLSX into plot variables in python.        |
| 2      | Study various data visualization library of python such as Matplotlib, Seaborn, plotly etc.                              |
| 3      | Use standard datasets and draw Scatter plot, line chart, bar chart, histogram, heatmap, using different python libraries |
| 4      | Use different data visualization techniques to filter the data.  |
| 5      | Use different data visualization techniques to transform the data.   |
| 6      | Use multiple data source to draw various visualization patterns.   |
| 7      | Create a Time Series visualization For a sales dataset.  |
| 8      | Create a trend line with a confidence band in any suitable dataset.  |
| 9      | Show an example of Skewed data and removal of skewedness using data visualization Techniques.                            |
| Sugge  | sted Books:  |

- Visualization Analysis & Design by Tamara Munzner (2014) (ISBN 9781466508910)
- Interactive Data Visualization for the Web by Scott Murray 2nd Edition (2017)





## OFFICE OF THE DEAN ACADEMICS

| VI Semester B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
|---|-------------------------|--|
| 6AM4-01: Compiler Design  |                         |  |
| Credit: 3 Max. Marks: 100 ( IA:30, ETE:70)                        |                         |  |
| 3L+0T+ 0P   | End Term Exams: 3 Hours |  |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- Familiar with basic ideas and the working of the compiler.
- Learn about syntax analysis.
- Learn about representation in the form of DAG.
- Learn about theory knowledge of Parsing, Code generation, and optimization.

#### **Course Outcomes:**

Upon successful completion of the course, students will be able to

- **CO-1:** Acquire knowledge of different phases and passes of the compiler and use compiler tools like LEX and YACC.
- **CO-2:** Understand the Top-Down and Bottom-up parsers and construction of LL, SLR, CLR, and LALR parsing tables.
- **CO-3:** Acquire knowledge about runtime data structure, like symbol table organization and different techniques.
- **CO-4**: Understand the target machine's run time environment, its instruction set for code generation, and techniques for code optimization.

| S. No. | Contents  | Hours |
|--------|---|-------|
| 1      | Introduction: Objective, scope, and outcome of the course. Compiler, Translator, Interpreter  | 6     |
|        | definition, Phase of compiler, Bootstrapping, Review of Finite automata lexical analyzer,     |       |
|        | Input, Recognition of tokens, Idea about LEX: A lexical analyzer generator, Error handling.   |       |
| 2      | Review of CFG Ambiguity of grammars: Introduction to parsing. Top-down parsing, LL            | 10    |
|        | grammars & passers error handling of LL parser, Recursive descent parsing predictive          |       |
|        | parsers, Bottom-up parsing, Shift reduce parsing, LR parsers, Construction of SLR, Conical    |       |
|        | LR & LALR parsing tables, parsing with ambiguous grammar. Operator precedence parsing,        |       |
|        | Introduction of automatic parser generator: YACC error handling in LR parsers.                |       |
| 3      | Syntax-directed translation: Construction of syntax trees, S-Attributed Definition, L-        | 10    |
|        | attributed definitions, Top-down translation. Intermediate code forms using postfix notation, |       |
|        | DAG, Three address code, TAC for various control structures, Representing TAC using           |       |
|        | triples and quadruples, Boolean expression, and control structures.                           |       |
| 4      | Runtime environments: Storage allocation, Strategies, heap management, Activation             | 8     |
|        | records, Accessing local and non-local names in a block structured language, Parameters       |       |
|        | passing, Symbol table organization, Data structures used in symbol tables.                    |       |
| 5      | <b>Definition of basic block control flow graphs:</b> DAG representation of basic block,      | 6     |
|        | Advantages of DAG, Sources of optimization, Loop optimization, Loop invariant                 |       |
|        | computation, Peephole optimization, Issues in the design of code generator, A simple code     |       |
|        | generator, Code generation from DAG. Machine Independent Optimization: Idea about             |       |
|        | global data flow analysis, constant propagation, liveness analysis, and common                |       |
|        | subexpression elimination.  |       |
| Total  |   |       |





## OFFICE OF THE DEAN ACADEMICS

- 1. Compilers: Principles, Techniques, and Tools, Second Edition, Alfred Aho, Monica Lam, Ravi Sethi, Jeffrey D. Ullman, January 2013. ISBN-978-9332518667.
- 2. Modern Compiler Implementation in Java. Andrew W Appel, Jens Paisberg. Cambridge University Press, January 2002. ISBN-978-0521820608
- 3. Modern Compiler Implementation in ML, Andrew W Appel, Cambridge University Press, December 1997. ISBN-0 521 58274 1
- 4. Modern Compiler Implementation in C, Andrew W Appel, Cambridge University Press, December 1997. ISBN 0-521-60765-5
- 5. Compiler Construction: Principles and Practice, 1st Edition, Kenneth C. Louden, Cengage Learning; 1 edition (January 24, 1997), ISBN-13: 978-0534939724
- 6. V Raghvan, "Principles of Compiler Design," McGraw-Hill, ISBN:9780070144712





## OFFICE OF THE DEAN ACADEMICS

| VI Semester   |                         |  |
|---|-------------------------|--|
| B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
| 6AM4-02: Design and Analysis of Algorithms            |                         |  |
| Credit: 3 Max. Marks: 100 ( IA:30, ETE:70 )           |                         |  |
| 3L+0T+ 0P   | End Term Exams: 3 Hours |  |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- Able to analyze asymptotic runtime complexity of algorithms including formulating recurrence relations.
- Able to understand and design algorithms using greedy strategy, divide and conquer approach, dynamic programming.
- Demonstrate a familiarity with major algorithms and data structures and Synthesize efficient algorithms in common engineering design situations

#### **Course Outcomes:**

Upon successful completion of the course the students will be able to

- CO-1: The ability of how to design an algorithm which solves the current problem in hand.
- **CO-2:** To Write efficient algorithms for given problems.
- **CO-3:** To focus on Deriving the complexities of any given algorithm.

CO-4: Learning the programming of various algorithms through assignments

| S. No. | Contents   |    |  |
|--------|--|----|--|
| 1      | Introduction: Concept of algorithmic efficiency, run time analysis of algorithms,              | 5  |  |
|        | Asymptotic Notations. Growth of Functions, Master's Theorem,                                   |    |  |
| 2      | Searching and Sorting: Structure of divide-and-conquer algorithms; examples: binary            |    |  |
|        | search, quick sort, Strassen Matrix Multiplication; merge sort, heap sort and Analysis of      |    |  |
|        | divide and conquer run time, recurrence relations.   |    |  |
| 3      | <b>Greedy Method:</b> Overview of the greedy paradigm examples of exact optimization solution: | 8  |  |
|        | minimum cost spanning tree, approximate solutions: Knapsack problem, Kruskal's algorithm       |    |  |
|        | and Prim's algorithm for finding Minimum cost Spanning Trees, Dijkstra's and Bellman           |    |  |
|        | Ford Algorithm for finding Single source shortest paths, Huffman coding, Activity Selection    |    |  |
|        | Problem.   |    |  |
| 4      | Dynamic programming: Principles of dynamic programming. Applications: Rod cutting              | 7  |  |
|        | problem, Floyd-Warshall algorithm for all pair shortest paths. Matrix multiplication,          |    |  |
|        | travelling salesman Problem, Longest Common sequence, Back tracking: Overview, 8-queen         |    |  |
|        | problem, and Knapsack problem, Traveling Salesman problem.                                     |    |  |
| 5      | Branch and bound: LC searching Bounding, FIFO branch and bound, LC branch and bound            | 6  |  |
|        | application: 0/1 Knapsack problem  |    |  |
| 6      | Computational Complexity: Polynomial Vs non-polynomial time complexity; NP-hard and            | 7  |  |
|        | NP-complete classes, examples: Circuit Satisfiablity, Vertex cover, Subset Sum problem,        |    |  |
|        | Randomized Algorithms, String Matching, NP-Hard and NP Completeness, Introduction to           |    |  |
|        | Approximation Algorithms,  |    |  |
|        | Total  | 40 |  |

- 1. T.H. Cormen, C.E. Leiserson, R.L. Rivest "Introduction to Algorithms", 3rd Ed., PHI, 2011 (reprint)
- 2. E. Horowitz, S. Sahni, and S. Rajsekaran, "Fundamentals of Computer Algorithms," Galgotia Publication
- 3. Sara Basse, A. V. Gelder, "Computer Algorithms," Addison Wesley
- 4. Aho ,Ullman "Principles of Algorithms"
- 5. S.K Basu- Design Methods and Analysis of Algorithms, 2nd Ed., PHI





## OFFICE OF THE DEAN ACADEMICS

| VI Semester B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
|---|-------------------------|--|
| 6AM4-03: Big Data Analytics                                       |                         |  |
| Credit: 3 Max. Marks: 100 ( IA:30, ETE:70 )                       |                         |  |
| 3L+0T+ 0P   | End Term Exams: 3 Hours |  |

#### **Course Objectives:**

As a result of successfully completing this course, students will:

- To understand the need of Big Data, challenges and different analytical architectures
- Installation and understanding of Hadoop Architecture and its ecosystems
- Processing of Big Data with Advanced architectures like Pig, Hive.

### **Course Outcomes:**

Upon successful completion of the course, students will be able to

- CO-1: Discuss the challenges and their solutions in Big Data
- **CO-2:** Understand and work on Hadoop Framework and eco systems.
- CO-3: Analyze the Big Data using Map-reduce programming in Hadoop
- CO-4: Applying Structure to Hadoop Data with Hive

| S. No. | Contents  | Hours |
|--------|---|-------|
| 1      | Introduction: Objective, scope and outcome of the course.                                   | 1     |
| 2      | Introduction to Big Data: Big data features and challenges, Problems with Traditional       | 7     |
|        | Large-Scale System, Sources of Big Data, 3 V's of Big Data, Types of Data.                  |       |
|        | Working with Big Data: Google File System. Hadoop Distributed File System (HDFS) -          |       |
|        | Building blocks of Hadoop (Namenode. Data node. Secondary Namenode. Job Tracker. Task       |       |
|        | Tracker), Introducing and Configuring Hadoop cluster (Local. Pseudo-                        |       |
|        | distributed mode, Fully Distributed mode). Configuring XML files.                           |       |
| 3      | Writing MapReduce Programs: A Weather Dataset. Understanding Hadoop API for                 | 8     |
|        | MapReduce Framework (Old and New). Basic programs of Hadoop MapReduce: Driver               |       |
|        | code. Mapper code, Reducer code. Record Reader, Combiner, Partitioner.                      |       |
| 4      | Hadoop I/O: The Writable Interface. Writable Comparable and comparators. Writable           | 8     |
|        | Classes: Writable wrappers for Java primitives. Text. Bytes Writable. Null Writable, Object |       |
|        | Writable and Generic Writable. Writable collections. Implementing a Custom Writable:        |       |
|        | Implementing a Raw Comparator for speed, Custom comparators.                                |       |
| 5      | Pig: Hadoop Programming Made Easier Admiring the Pig Architecture, Going with the Pig       | 8     |
|        | Latin Application Flow. Working through the ABCs of Pig Latin. Evaluating Local and         |       |
|        | Distributed Modes of Running Pig Scripts, Checking out the Pig Script Interfaces, Scripting |       |
|        | with Pig Latin.   |       |
| 6      |   |       |
|        | Hive is Put Together, Getting Started with Apache Hive. Examining the Hive Clients.         |       |
|        | Working with Hive Data Types. Creating and Managing Databases and Tables, Seeing How        |       |
|        | the Hive Data Manipulation Language Works, Querying and Analyzing Data.                     |       |
|        | Total   | 40    |

- 1. Mike Frampton, "Mastering Apache Spark", Packt Publishing, 2015.
- 2. Tom White, "Hadoop: The Definitive Guide", O'Reilly, 4th Edition, 2015.
- 3. Nick Pentreath, "Machine Learning with Spark", Packt Publishing, 2015



# BIKANER TECHNICAL UNIVERSITY, BIKANER बीकानेर तकनीकी विश्वविद्यालय, बीकानेर OFFICE OF THE DEAN ACADEMICS



- 4. Mohammed Guller, Big Data Analytics with Spark, Apress, 2015
- 5. Donald Miner, Adam Shook, "Map Reduce Design Pattern", O'Reilly, 2012





## OFFICE OF THE DEAN ACADEMICS

| VI Semester   |                         |  |
|---|-------------------------|--|
| B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
| 6AM4-04: Digital Image Processing                     |                         |  |
| Credit: 3 Max. Marks: 100 ( IA:30, ETE:7              |                         |  |
| 3L+0T+ 0P   | End Term Exams: 3 Hours |  |

**Course Objectives**: As a result of successfully completing this course, students will:

- To learn the fundamental concepts of Digital Image Processing.
- Able to Understand basic image processing operations.
- To understand image analysis algorithms.
- Exposure to current applications in the field of digital image processing.

Course Outcomes: Upon successful completion of the course, students will be able to

- **CO-1:** Review the fundamental concepts of digital image processing systems.
- **CO-2:** Analyze images in the frequency domain using various transforms.
- **CO-3:** Evaluate the techniques for image enhancement, image restoration, and Morphological Operation.
- CO-4: Categorize various compression techniques.
- **CO-5**: Interpret image segmentation and representation techniques.

| S. No. | Contents  |    |
|--------|---|----|
| 1      | Introduction: Objective, scope and outcome of the course.   |    |
| 2      | <b>Introduction to Image Processing:</b> Digital Image representation, Sampling & Quantization, Steps in image Processing, Image acquisition, color image representation.   |    |
| 3      | <b>Image Transformation &amp; Filtering:</b> Intensity transform functions, histogram processing, Spatial filtering, Fourier transforms and its properties, frequency domain filters, colour models, Pseudo colouring, colour transforms, Basics of Wavelet Transforms. |    |
| 4      | <b>Image Restoration:</b> Image degradation and restoration process, Noise Models, Noise Filters, degradation function, Inverse Filtering, Homomorphism Filtering.  |    |
| 5      |   |    |
| 6      | <b>Image Segmentation &amp; Representation:</b> Point, Line and Edge Detection, Thresholding, Edge and Boundary linking, Hough transforms, Region Based Segmentation, Boundary representation, Boundary Descriptors.  |    |
|        | Total   | 40 |

- 1. Gonzalez C. R., Woods E. R., Digital Image Processing, Pearson Education (2008) 3rd ed.
- 2. A.K.Jain, "Fundamentals of Digital Image Processing", PHI,1995
- 3. Sonka M., Hlavac V. and Boyle R., Image Processing, Analysis and Machine Vision, Thomson Learning, (1993)1st ed.
- 4. McAndrew A., Introduction to Digital Image Processing with Matlab, Thomson Course Technology (2004)
- 5. Low A., Introductory Computer Vision and Image Processing, McGraw-Hill (1991), 1st ed.
- 6. Boyle and Thomas: Computer Vision A First Gurse 2nd Edition, ISBN 0-632-028-67X, Blackwell Science 1995.
- 7. Pakhera Malay K: Digital Image Processing and Pattern Recognation, PHI.





## OFFICE OF THE DEAN ACADEMICS

| VI Semester B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
|---|-------------------------|--|
| 6AM4-05: Deep Learning  |                         |  |
| Credit: 3 Max. Marks: 100 ( IA:30, ETE:70)                        |                         |  |
| 3L+0T+ 0P   | End Term Exams: 3 Hours |  |

**Course Objectives**: As a result of successfully completing this course, students will:

- To describe the major differences between deep learning and other types of machine learning algorithms.
- To explain the fundamental methods involved in deep learning.
- To understand various aspects of Deep Earning and its building block.
- To understand and differentiate between the major types of neural network architectures.
- To Select or design neural network architectures for new data problems based on their requirements and problem characteristics and analyze their performance.
- To understand basic working principles and how Deep Learning is used to solve real-world problems

Course Outcomes: Upon successful completion of the course, students will be able to

- **CO-1:** Able to learn the fundamental concepts of neural networks and deep neural networks.
- **CO-2:** Able to understand the working principle of convolution neural networks.
- **CO-3:** Able to perform hyperparameter tuning.
- **CO-4**: Able to analyze and design neural network for real work problem.
- CO-5: Able to understand working principle of various types of neural networks

| CO-5:   | : Able to understand working principle of various types of neural networks.            |       |
|---|--|-------|
| S. No.  | Contents   | Hours |
| 1   | Introduction to Neural Networks  | 7     |
|   | Introduction of artificial neural network and deep learning, characteristics of neural |       |
|   | networks terminology, neurons, perceptron, backpropagation, Basic learning laws,       |       |
|   | Activation and Loss function - Function approximation, applications                    |       |
| 2 Introduction to Convolution Neural Networks |  | 9     |
|   | CNN Architecture and Operations, convolutional layer, Pooling layer, Variants of the   |       |
|   | Convolution Model, Forward and Backward propagation, Building a Deep Neural            |       |
|   | Network  |       |
|   | Improving Deep Neural Networks   |       |
|   | Training a deep neural network, hyper-parameter tuning, Hidden layers, Generalization  |       |
|   | Gap – Under-fitting Vs Over-fitting – Optimization, Normalization                      |       |
| 3   | 3 Practical aspects of Deep Learning: Train/Dev / Test sets, Bias/variance, Overfi     |       |
|   | and regularization, Linear models and optimization, Vanishing/exploding gradients,     |       |
|   | Gradient checking - Logistic Regression, Convolution Neural Networks, RNN and          |       |
|   | Backpropagation – Convolutions and Pooling.  |       |
| 4   | Optimization algorithms: Mini-batch gradient descent, exponentially weighted           | 8     |
|   | averages, RMS prop, Learning rate decay, the problem of local optima, Batch norm –     |       |
|   | Parameter tuning process.  |       |
| 5   | Neural Network Architectures: Recurrent Neural Networks, Adversarial NN, Spectral      | 9     |
|   | CNN, Self-Organizing Maps, Restricted Boltzmann Machines, Long Short-Term Memory       |       |
|   | Networks (LSTM) and Deep Reinforcement Learning – Tensor Flow, Keras or MatConvNet     |       |
|   | for implementation.  |       |
|   | Total  | 42    |
| Sugges  | sted Books:  |       |
| 1.  |  |       |
| -   |  |       |





## OFFICE OF THE DEAN ACADEMICS

- https://www.deeplearningbook.org/)
- 2. Deep Learning Step by Step with Python, N D Lewis, 2016
- 3. Deep Learning: A Practitioner's Approach, Josh Patterson, Adam Gibson, O'Reilly Media, 2017
- 4. Deep Learning, Ian Goodfellow Yoshua Bengio Aaron Courville, MIT Press, 2017
- 5. James Allen "Natural Language Understanding", Pearson Publication 8th Edition. 2012.
- 6. François Chollet "Deep Learning with Python," First Edition, Manning Publication, 2018
- 7. Neural Networks and Deep Learning, Michael Nielsen, Determination Press (2015) (link: http://neuralnetworksanddeeplearning.com/)





## OFFICE OF THE DEAN ACADEMICS

| VI Semester B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
|---|-------------------------|--|
| 6AM5-11: Internet of Things                                       |                         |  |
| Credit:2 Max. Marks: 100 ( IA:30, ETE:70                          |                         |  |
| 2L+0T+ 0P   | End Term Exams: 3 Hours |  |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- Able to Understand the fundamentals about IoT
- Able to Understand about IoT Access technologies
- Able to Understand the design methodology and different IoT hardware platforms.
- Able to Understand the basics of IoT Data Analytics and supporting services.
- Able to Understand about various IoT case studies and industrial applications.

### **Course Outcomes:**

Upon successful completion of the course, students will be able to

- CO-1: Understand the basics and Architecture of IoT
- CO-2: Understand design methodology and hardware platforms involved in IoT
- CO-3: Analyze the challenges in IoT based design and development
- CO-4: Understand IOT Applications in Industrial & real world.

| S. No. | Contents  | Hours |
|--------|---|-------|
| 1      | Introduction: Objective, scope and outcome of the course.   |       |
| 2      | Introduction to IoT: Definition and characteristics of IoT, Design of IOT: Physical design of IOT, Logical Design of IOT- Functional Blocks, communication models, communication APIs, IOT enabling Technologies- Wireless Sensor Networks, Cloud computing, big data analytics, embedded systems. IOT Levels and deployment templates. |       |
| 3      | IoT Hardware and Software: Sensor and actuator, Humidity sensors, Ultrasonic sensor, Temperature Sensor, Arduino, Raspberry Pi, LiteOS, RIoTOS, Contiki OS, Tiny OS.  | 7     |
| 4      | Architecture and Reference Model: Introduction, Reference Model and architecture, Representational State Transfer (REST) architectural style, Uniform Resource Identifiers (URIs). Challenges in IoT- Design challenges, Development challenges, Security challenges, Other challenges.   |       |
| 5      | IOT and M2M: M2M, Difference and similarities between IOT and M2M, Software defined networks, network function virtualization, difference between SDN and NFV for IoT. Case study of IoT Applications   | 7     |
|        | Total   | 28    |

- 1. IoT Fundamentals: Networking Technologies, Protocols and Use Cases for Internet of Things, David Hanes, Gonzalo Salgueiro, Patrick Grossetete, Rob Barton and Jerome Henry, Cisco Press, 2017
- 2. Internet of Things A hands-on approach, Arshdeep Bahga, Vijay Madisetti, Universities Press, 2015
- 3. Internet of Things: Architecture, Design Principles And Applications, Rajkamal, McGraw Hill HigherEducation
- 4. "From Machine-to-Machine to the Internet of Things Introduction to a New Age of Intelligence" Jan Höller, Vlasios Tsiatsis, Catherine Mulligan, Stamatis Karnouskos, Stefan Avesand, David Boyle, Elsevier, 2014.





## OFFICE OF THE DEAN ACADEMICS

| VI Semester B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
|---|-------------------------|--|
| 6AM5-12: Pattern Recognition                                      |                         |  |
| Credit:2 Max. Marks: 100 ( IA:30, ETE:70                          |                         |  |
| 2L+0T+ 0P   | End Term Exams: 3 Hours |  |

#### **Course Objectives:**

As a result of successfully completing this course, students will:

• Students should be able to understand soft computing concepts and techniques and foster their abilities in designing and implementing soft computing based solutions for real-world problems.

#### **Course Outcomes:**

Upon successful completion of the course, students will be able to

**CO-1:** Describe and compare a variety of pattern classification, structural pattern recognition, and pattern classifier combination techniques

**CO-2:** Apply pattern recognition techniques to real-world problems such as document analysis and recognition

CO-3: Summarize, analyze and relate research in the pattern recognition area

| S. No. | Contents   | Hours |  |
|--------|--|-------|--|
| 1      | Introduction: Objective, scope and outcome of the course.                                    |       |  |
| 2      | Basics Of Probability, Random Processes And Linear Algebra, Bayes Decision                   |       |  |
|        | <b>Theory</b> : Bayes' theorem, Minimum-error-rate classification, Classifiers, Discriminant |       |  |
|        | functions, Decision surfaces, Normal density and discriminant functions, Discrete            |       |  |
|        | Features   |       |  |
| 3      | Parameter Estimation Methods: Maximum-Likelihood estimation, Gaussian case,                  |       |  |
|        | Maximum a Posteriori estimation, Bayesian estimation, Gaussian case                          |       |  |
| 4      | Unsupervised Learning and Clustering: Criterion functions for clustering, Algorithms         |       |  |
|        | for clustering, K-Means, Hierarchical and other methods, Cluster validation, Gaussian        |       |  |
|        | mixture models, Expectation-Maximization method for parameter estimation, Maximum            |       |  |
|        | entropy estimation   |       |  |
| 5      | Sequential Pattern Recognition: Hidden Markov Models (HMMs), Discrete Hmms,                  | 7     |  |
|        | Continuous HMMs Nonparametric Techniques For Density Estimation Parzen-Window                |       |  |
|        | Method, K-Nearest Neighbor Method  |       |  |
|        | Total  | 28    |  |

- 1. Pattern Classification, Richard O. Duda, Peter E. Hart, David G. Stork John Wiley 2001
- 2. Pattern Recognition, Konstantinos Koutroumbas and Sergios Theodoridis 4th Edition., Academic Press 2009
- 3. Pattern Recognition and Machine Learning, Bishop, Christopher, Springer 2006





## OFFICE OF THE DEAN ACADEMICS

| VI Semester   |                         |  |
|---|-------------------------|--|
| B. Tech. (Artificial Intelligence & Machine Learning) |                         |  |
| 6AM5-13: GPU Computing                                |                         |  |
| Credit:2 Max. Marks: 100 (IA:30, ETE:70 )             |                         |  |
| 2L+0T+ 0P   | End Term Exams: 3 Hours |  |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- Understand parallel programming with graphics processing units (GPUs).
- Understand Memory management and mechanism for parallel computing

#### **Course Outcomes:**

Upon successful completion of the course, students will be able to

- **CO-1:** Define and understand terminology commonly used in parallel computing.
- CO-2: Describe common GPU architectures and programming models.
- **CO-3:** Understand a Given problem and develop an efficient parallel algorithm to solve it.

**CO-4:** Understand CUDA memory access mechanism.

| S. No. | Contents   | Hours |
|--------|--|-------|
| 1      | <b>Introduction:</b> Objective, scope and outcome of the course.   | 1     |
| 2      | <b>GPU Introduction</b> : To study architecture and capabilities of modern GPUs and learn programming techniques for the GPU such as CUDA programming model.   | 6     |
|        | Heterogeneous Parallel Computing, Architecture of a Modern GPU, Speeding Up Real Applications, Parallel Programming Languages and Models.  |       |
| 3      | <b>History of GPU Computing</b> : Evolution of Graphics Pipelines, The Era of Fixed-Function Graphics Pipelines, Evolution of Programmable Real-Time Graphics, Unified Graphics and Computing Processors, GPGPU, Scalable GPUs, Recent Developments, Future Trends.                | 5     |
| 4      | <b>Introduction to Data Parallelism and CUDA C:</b> Data Parallelism, CUDA Program Structure, A Vector Addition Kernel, Device Global Memory and Data Transfer, Kernel Functions and Threading.  | 5     |
| 5      | <b>Data-Parallel Execution Model:</b> CUDA Thread Organization, Mapping Threads to Multidimensional Data, Matrix-Matrix Multiplication—A More Complex Kernel, Synchronization and Transparent Scalability, Assigning Resources to Blocks, Thread Scheduling and Latency Tolerance. | 6     |
| 6      | <b>CUDA Memories:</b> Importance of Memory Access Efficiency, CUDA Device Memory Types, A Tiled Matrix – À Matrix Multiplication Kernel, Memory as a Limiting Factor to Parallelism.   | 5     |
|        | Total  | 28    |

- 1. Sanders, J. and Kandrot, E., CUDA by Example: An Introduction to General-Purpose GPU Programming, Addison-Wesley Professional (2012) 4th Edition.
- 2. Kirk, D. and Hwu, M., W., Programming Massively Parallel Processors: A Hands-on Approach. Morgan Kaufmann (2016) 3rd Edition.
- 3. Hwu, M., W., A GPU Computing Gems Emerald Edition (Applications of GPU Computing Series), Morgan Kaufmann (2011) 1st Edition.





## OFFICE OF THE DEAN ACADEMICS

|        | D Took (Autitional)   | VI Semester<br>Intelligence & Machine Learning)  |
|--------|---|--|
|        |   | and Analysis of Algorithms Lab   |
| Credit |   | Max. Marks: 100 ( IA:60, ETE:40 )  |
| 0L+0T  | T+ 2P   | End Term Exams: 2 Hours  |
| Course |   | completing this course, students will: In the design and analysis of the major classes of algorithms a given computational task and to compare and contrast their  |
|        | e Outcomes: Upon successful completion  |  |
|        | Design algorithms using divide and conq   |  |
|        |   | ing, graph related and combinatorial algorithm in a high level   |
|        | language.   |  |
|        |   | and quick sort algorithms using divide and conquer technique.  |
|        | 11 0 0  | que to solve real world problems such as knapsack and TSP  |
| S. No. | L   | ist of Experiments   |
| 1      | elements. Repeat the experiment for diff  | uicksort method and determine the time required to sort the ferent values of n, the number of elements in the list to be a versus n. The elements can be read from a file or can be the terrator.              |
| 2      | required to sort the elements. Repeat the   | gorithm to sort a given set of elements and determine the time experiment for different values of n, the number of elements of the time taken versus n. The elements can be read from a file number generator. |
| 3      | a. Obtain the Topological ordering of ve<br>a given directed graph using Warshall's | ertices in a given digraph. b. Compute the transitive closure of algorithm.  |
| 4      | Implement 0/1 Knapsack problem using  | Dynamic Programming.   |
| 5      | From a given vertex in a weighted connection Dijkstra's algorithm.                  | ected graph, find shortest paths to other vertices using   |
| 6      | Find Minimum Cost Spanning Tree of a  | given undirected graph using Kruskal's algorithm.  |
| 7      | a. Print all the nodes reachable from a graph is connected or n                     | iven starting node in a digraph using BFS method. b. Check ot using DFS method.  |
| 8      | Find Minimum Cost Spanning Tree of a  | given undirected graph using Prim's algorithm.   |
| 1.T .H |   | "Introduction to Algorithms", 3rd Ed.,PHI, 2011 (reprint) undamentals of Computer Algorithms,"Galgotia Publication   |

5.S.K Basu- Design Methods and Analysis of Algorithms, 2nd Ed., PHI





## OFFICE OF THE DEAN ACADEMICS

| VI Semester B. Tech. (Artificial Intelligence & Machine Learning) |                                   |  |
|---|-----------------------------------|--|
| 6AM4-22: Big Data Analytics Lab                                   |                                   |  |
| Credit: 1   | Max. Marks: 100 ( IA:60, ETE:40 ) |  |
| 0L+0T+ 2P   | End Term Exams: 2 Hours           |  |

Course Objectives: As a result of successfully completing this course, students will:

- Perform data gathering of large data from a range of data sources.
- Critically analyze existing Big Data datasets and implementations, taking practicality, and usefulness metrics into consideration.
- Able to select and apply suitable statistical measures ,analyses techniques for data of various structure and content and present summary statistics

Course Outcomes: Upon successful completion of the course, students will be able to

- **CO-1:** Describe the key issues in Big Data Management and experiment with the Hadoop framework.
- **CO-2:** Explain the structure and unstructured data by using NoSQL commands.
- **CO-3:** Apply scientific computing algorithms for finding similar items and clustering.
- CO-4: Test fundamental enabling techniques and scalable algorithms for data stream mining.
- **CO-5:** Develop problem solving and critical thinking skills in fundamental enable techniques like Hadoop & MapReduce.

| List of Experiments   |
|---|
| Implement the following Data structures in Java   |
| i) Linked Lists ii) Stacks iii) Queues iv) Set v) Map   |
| Perform setting up and Installing Hadoop in its three operating modes:                                |
| Standalone, Pseudo distributed, Fully distributed.  |
| Implement the following file management tasks in Hadoop:  |
| • Adding files and directories • Retrieving files   |
| • Deleting files Hint: A typical Hadoop workflow creates data files (such as log files) elsewhere     |
| and copies them into HDFS using one of the above command line utilities.                              |
| Run a basic Word Count Map Reduce program to understand Map Reduce Paradigm.                          |
| Write a Map Reduce program that mines weather data. Weather sensors collecting data every             |
| hour at many locations across the globe gather a large volume of log data, which is a good            |
| candidate for analysis with MapReduce, since it is semi structured and record-oriented.               |
| Implement Matrix Multiplication with Hadoop Map Reduce  |
| Install and Run Pig then write Pig Latin scripts to sort, group, join, project, and filter your data. |
| Install and Run Hive then use Hive to create, alter, and drop databases, tables, views, functions,    |
| and indexes.  |
| Solve some real life big data problems.   |
|   |

- 1. Mike Frampton, "Mastering Apache Spark", Packt Publishing, 2015.
- 2. Tom White, "Hadoop: The Definitive Guide", O'Reilly, 4th Edition, 2015.
- 3. Nick Pentreath, "Machine Learning with Spark", Packt Publishing, 2015
- 4. Mohammed Guller, Big Data Analytics with Spark, Apress, 2015
- 5. Donald Miner, Adam Shook, "Map Reduce Design Pattern", O'Reilly, 2012





## OFFICE OF THE DEAN ACADEMICS

| VI Semester B. Tech. (Artificial Intelligence & Machine Learning) |                                   |  |
|---|-----------------------------------|--|
| 6AM4-23: Digital Image Processing Lab                             |                                   |  |
| Credit: 1   | Max. Marks: 100 ( IA:60, ETE:40 ) |  |
| 0L+0T+ 2P   | End Term Exams: 2 Hours           |  |

### **Course Objectives:**

As a result of successfully completing this course, students will:

- To introduce the concepts of image processing and basic analytical methods to be used in image processing.
- To familiarize students with image enhancement and restoration techniques
- To explain different image compression techniques. To introduce segmentation and morphological processing techniques.

#### **Course Outcomes:**

Upon successful completion of the course, students will be able to

- **CO-1:** Review the fundamental concepts of a digital image processing system.
- **CO-2:** Analyze images in geometric transforms with image rotation, scaling, and translation.
- **CO-3:** Evaluate the techniques for image enhancement and image restoration.
- CO-4: Categorize various compression techniques and Interpret Image compression standards
- **CO-5:** Interpret image segmentation and representation techniques.

| S. No. | List of Experiments  |
|--------|--|
| 1      | Point-to-point transformation. This laboratory experiment provides for thresholding an image and the   |
|        | evaluation of its histogram. Histogram equalization. This experiment illustrates the relationship      |
|        | among the intensities (gray levels) of an image and its histogram.                                     |
| 2      | Geometric transformations. This experiment shows image rotation, scaling, and translation. Two-        |
|        | dimensional Fourier transform  |
| 3      | Linear filtering using convolution. Highly selective filters.  |
| 4      | Ideal filters in the frequency domain. Non Linear filtering using convolutional masks. Edge detection. |
|        | This experiment enables students to understand the concept of edge detectors and their operation in    |
|        | noisy images.  |
| 5      | Morphological operations: This experiment is intended so students can appreciate the effect of         |
|        | morphological operations using a small structuring element on simple binary images. The operations     |
|        | that can be performed are erosion, dilation, opening, closing, open-close, close-open.                 |

- 1. Digital Image Processing, Rafea l C. Gonzalez, Richard E. Woods, Second Edition, Pearson Education/PHI
- 2. Image Processing, Analysis, and Machine Vision, Milan Sonka, Vaclav Hlavac and Roger Boyle, Second Edition, Thomson Learning.
- 3. Digital Image Processing using Matlab, Rafeal C. Gonzalez, Richard E. Woods, Steven L. Eddins, Pearson Education.





## OFFICE OF THE DEAN ACADEMICS

| VI Semester B. Tech. (Artificial Intelligence & Machine Learning) |   |  |
|---|---|--|
| 6AM7-50: Innovation and Design Thinking Hands-on Project          |   |  |
| Credit: 2   | Max. Marks: 100 ( IA:60, ETE:40 )           |  |
| 0L+0T+3P  | Mode of evaluation: Report and presentation |  |

#### **Course Objectives:**

As a result of successfully completing this course, students will:

- Learn about the National Innovation and Startup Policy (NISP)of Govt. of India.
- Learn how to ideate, prototype and Iterate solutions.
- Learn about applying Design Thinking Tools and Approaches for Right Problem Identification and Solution Development.
- Learn about Business Plan Development.
- Learn about Legal Structures and Ethical Steps in Establishing Startups.
- Able to design and develop a Prototype.
- Students will be able to pitch their idea.
- Will be able to demonstrate their innovative and design thinking capabilities using mock-up models.

### **Course Outcomes:**

Upon successful completion of the course, students will be able to

- **CO-1:** learn about opportunities and challenges for startup and incubation.
- **CO-2:** Students will be able to identify an Opportunity from a Problem using design thinking.
- **CO-3:** Students will be able to frame Product and service ideas.
- **CO-4**: Learn and implement Design Thinking Process.
- **CO-5:** Students will be able to design and develop a Prototype.
- **CO-6**: Students will be able to prepare documentation and pitch their idea.

| exp. No. | Contents  |
|----------|---|
| 1        | National Innovation and Startup Policy (NISP) and Legal Structures and Ethical Steps in Establishing      |
|          | Startups, Generation and Management of IP at the Early Stage of Innovation and Startup Development,       |
|          | IPR and IPR policies.   |
| 2        | Design Thinking, Process of Design Thinking, Empathy, Define, Ideate, Prototype, Testing.                 |
| 3        | Understanding Technology Readiness Level (TRL), Manufacturing Readiness Level (MRL) and                   |
|          | Investment Readiness Level (IRL) Stages & Implications in Innovation Development                          |
| 4        | Capstone Project:   |
|          | Students in groups of 3 to 5 students must prepare a project idea using the design thinking process under |
|          | the mentorship of the faculty members. Students must submit a capstone project report containing          |
|          | various ideas learned in experiments numbers 1-3 and their implementation or usage in the capstone        |
|          | project to the Institute Innovation Council (IIC) cell or Head of Department along with a presentation.   |

#### **Assessment or Evaluation:**

Students need to submit a capstone project report to the Institute Innovation Council (For the Institute having IIC cells) or the head of the department (For the Institute not having IIC cells) containing step by step approach to the project based on design thinking methodology along with the final presentation to IIC Cell (For the Institute having IIC cells) or Head of department (For the Institute not having IIC cells).

#### Suggested Books:

1. Idris Mootee, "Design Thinking for Strategic Innovation: What They Can't Teach You at Business or





## OFFICE OF THE DEAN ACADEMICS

Design School", John Wiley & Sons (2013).

- 2. Tim Brown, "Change by design", Harper Collins, 2009
- 3. "Design Thinking- The Guide Book" Facilitated by the Royal Civil service Commission, Bhutan
- 4. The Lean Startup: How Constant Innovation Creates Radically Successful Businesses
- 5. Start With Why: How Great Leaders Inspire Every
- 6. National Innovation and Startup Policy 2019 for students and faculty of Higher Education Institutions (HEIs) https://mic.gov.in/assets/doc/startup\_policy\_2019.pdf
- 7. Tom Kelley, The Art of Innovation: Lessons in Creativity from IDEO, America's Leading Design Firm
- 8. Roger L. Martin, Design of Business: Why Design Thinking is the Next Competitive Advantage, Harvard Business Review Press
- 9. Online resource