

NATIONAL INSTITUTE OF ELECTRONICS AND INFORMATION TECHNOLOGY

DEEMED TO BE UNIVERSITY UNDER DISTINCT CATEGORY
(AN AUTONOMOUS INSTITUTION UNDER MINISTRY OF ELECTRONICS & IT, GOVT. OF INDIA)

Curriculum and Syllabi for M.Tech - Cyber Forensics (2026-27 Scheme)



Main Campus at Ropar and Constituent Units at Aizawl, Agartala, Aurangabad, Calicut, Gorakhpur, Imphal, Itanagar, Kekri, Kohima, Patna and Srinagar

Vision

To be a premier center of excellence in Cyber Forensics education, research, and innovation, fostering globally competent professionals who safeguard digital ecosystems through cutting-edge forensic techniques, ethical practices, and technological leadership.

Mission

The mission of the M.Tech in Cyber Forensics program is to equip students with advanced knowledge and practical expertise in digital forensics, cybersecurity, and investigative methodologies. The program aims to foster analytical thinking, innovation, and ethical responsibility to prepare graduates for critical roles in securing digital assets, conducting cyber investigations, and contributing to national and global cybersecurity efforts.

Program Education Objectives

PEO1: Attain in-depth expertise in cyber forensics, digital investigation, and cybersecurity, grounded in strong theoretical and practical foundations.

PEO2: Excel in careers in law enforcement, intelligence, private sector cybersecurity, and academic research by addressing complex cybercrime and digital evidence challenges.

PEO3: Demonstrate leadership, collaboration, and communication skills while contributing to technological advancements and policy formulation in cyber law and digital security.

PEO4: Adapt to emerging trends, legal frameworks, and technologies in the evolving landscape of digital forensics, cybersecurity, and privacy protection.

Program Outcomes

PO1: Ability to independently conduct investigations and research in cyber forensics and digital crime analysis.

PO2: Ability to write, present, and publish substantial technical reports, forensic case documentation, and scholarly research in cyber forensics and cybersecurity domains.

PO3: Mastery of tools and techniques in cyber forensics, including evidence acquisition, digital artifact analysis, memory forensics, and malware analysis.

PO4: Proficiency in using forensic and cybersecurity tools such as EnCase, FTK, Autopsy, Wireshark, and Metasploit for practical case resolution.

PO5: Capacity for continuous learning, innovation, and ethical decision-making to handle emerging cyber threats, legal challenges, and privacy concerns.

Program Specific Outcomes

PSO1: Apply forensic methodologies to identify, preserve, analyze, and present digital evidence in alignment with legal and investigative standards.

PSO2: Design and implement secure systems and incident response strategies to detect and mitigate cyber threats in real-time.

PSO3: Analyze malware behavior, cyber attacks, and data breaches using automated tools and reverse engineering techniques.

PSO4: Demonstrate knowledge of cyber laws, digital ethics, and regulatory compliance in national and international contexts.

PSO5: Engage in advanced research and innovation in areas such as mobile forensics, IoT forensics, blockchain forensics, and AI-assisted threat intelligence.

CATEGORY WISE CREDIT DISTRIBUTION

Category	Category Code	Credits
Audit Courses (without grade or credit)	AU	0
Open Elective Courses	OE	8
Professional Elective Courses	PE	16
Program Core Courses	PC	20
Skill / Employment Enhancement Courses (Project/Internship/Seminar/Dissertation/Research)	EE	36
Total Credits (Common track)		80

EVALUATION AND ASSESSMENT

1. Performance of a student in a semester shall be evaluated through continuous Class assessment, Tutorial/Lab assessment, Mid-Semester Examination (MSE) and End-Semester Examination (ESE). Both the MSE and ESE shall be the University examination and will be conducted as notified by the Controller of Examinations (CoE) of the University.
2. The continuous assessment shall be based on assignments, tutorials, paper presentation / quizzes/ viva-voce / flipped classes, lab work / projects / fieldwork and attendance, etc.
3. The MSE/ESE shall be comprising of written papers.
4. The overall assessment of the students will be done as per the following scheme:

S. No.	Assessment Type	Marks Weightage
1.	Mid Semester Examination	20
2.	End Semester Examination	40
3.	Continuous Assessment - Practical /Lab/ Tutorial	25
4.	Continuous Assessment - Theory	15
Total		100

For more details, please refer the applicable ordinance for this programme and Assessment SOPs.

CURRICULUM

Semester 1

Semester Code	Course Code	Course Title	L	T	P	Cr
PC-MTECFS-101	DCSE250043	Cryptography and Network Security	3	0	2	4
PE-MTECFS-102	Elective	Program Elective				4
OE-MTECFS-103	Elective	Open Elective				4
PC-MTECFS-104	DASH250095	Research Methodology and IPR	3	1	0	4
AU-MTECFS-105	Elective	Audit Elective				0
PC-MTECFS-106	DCSE250065	Digital Forensics & Cyber Crime Investigation	3	0	2	4

Semester 2

Semester Code	Course Code	Course Title	L	T	P	Cr
PC-MTECFS-201	DCSA250021	Cyber Law	3	1	0	4
PC-MTECFS-202	DCSE250092	Mobile & Digital Forensic	3	0	2	4
PE-MTECFS-203	Elective	Program Elective				4
PE-MTECFS-204	Elective	Program Elective				4
AU-MTECFS-205	Elective	Audit Elective				0
EE-MTECFS-206	EE	Mini Project				4

Semester 3

Semester Code	Course Code	Course Title	L	T	P	Cr
PE-MTECFS-301	Elective	Program Elective				4
OE-MTECFS-302	Elective	Open Elective				4
EE-MTECFS-303	EE	Dissertation-I /Internal Project				12

Semester 4

Semester Code	Course Code	Course Title	L	T	P	Cr
EE-MTECFS-401	EE	Dissertation-II				20

Elective Courses

Elective Courses for PE Category

Course Code	Course Title	L	T	P	Cr	For Sem./Code
DOAI260001	AI (Industry)-Applied Ethics for AI	3	0	2	4	
DOAI260002	AI (Industry)-Introduction to Artificial Intelligence	3	0	2	4	
DOAI260003	AI (Industry)-Introduction to Generative AI	3	0	2	4	
DOAI260004	AI (Industry)-Introduction to Machine Learning	3	0	2	4	
DCSE250019	Cloud Architecture and Security	3	0	2	4	102
DCSE250093	Mobile & Wireless Security	3	0	2	4	102
DOAI250017	Data Analytics for Fraud Detection	3	0	2	4	102
DCSA250035	Information Security Audit	3	0	2	4	103
DCSE250049	Data Privacy	3	0	2	4	103
DCSE250072	Edge Computing	3	1	0	4	103
DCSA250005	Applied Cryptography	3	1	0	4	203
DOAI250032	Machine Learning	3	0	2	4	203
DOEE250124	IOT and its Applications	3	1	0	4	203
DCSA250008	Biometrics	3	1	0	4	301
DCSA250009	Blockchain Technology	3	1	0	4	301
DCSA250033	Image Forensics and Security	3	0	2	4	301

Elective Courses for OE Category

Students may opt courses under MOOC, NPTEL, SWAYAM, NSQF/NCVET aligned courses or other relevant technical subjects not included in their core curriculum. The exercise of option shall be subject to the Course Schedule of the respective Constituent Unit, the credit requirements and availability of seats. Guidelines for choosing MOOC/Online courses are given separately and shall have to be adhered to.

Elective Courses for AU (Audit) Category

Course Code	Course Title	L	T	P	Cr
DASH240027	Disaster Management	2	0	0	0
DASH240047	English for Research Paper Writing	2	0	0	0
DASH240052	Environmental Science	3	0	0	0
DASH240085	Pedagogy Studies	2	0	0	0
DASH240086	Personality Development through Life Enlightenment Skills	2	0	0	0
DASH240097	Sanskrit for Technical Knowledge	2	0	0	0
DASH240103	Stress Management by Yoga	2	0	0	0
DASH240104	Technical Presentation and Report Writing	0	0	2	0
DASH240106	Value Education	2	0	0	0
DASH240124	Constitution of India - AU	2	0	0	0
DASH250023	Constitution of India	2	1	0	0
DASH250027	Disaster Management	3	0	0	0

DASH250034	Energy and Environmental Engineering	3	0	0	0
DASH250047	English for Research Paper Writing	2	0	0	0
DASH250054	Essence of Indian Knowledge and Tradition	2	0	0	0
DASH250085	Pedagogy Studies	2	0	0	0
DASH250086	Personality Development	2	0	0	0
DASH250097	Sanskrit for Technical Knowledge	3	0	0	0
DASH250103	Stress Management by Yoga	3	0	0	0
DASH250106	Value Education	2	0	0	0
DASH250117	Innovative Thinking and Entrepreneurship Skills	1	0	0	0
DASH250118	Intellectual Property Rights	2	0	0	0
DCSA240080	Training on Computer Networking	0	0	2	0
DCSA240304	Lab Based on Statistical Package	0	0	2	0
Any other Scheduled Course as per the Course Schedule of the respective Constituent Unit can also be chosen, subject to availability of seats. However, there shall be no assessment and grading for the course chosen as Audit Course.					

The choice of all type of Electives available shall be as per the Course Schedule of the respective Constituent Unit.

Guidelines for Massive Open Online Courses (MOOC) / approved Online Courses

1. Relevant Swayam, MOOC, NPTEL, NIELIT NSQF and any other online courses approved by UGC/AICTE shall also be allowed as Open Electives(OE).
2. One credit of MOOC course should be minimum of 30 hours.
3. A student can choose any approved online course as Open Elective, subject to minimum credit requirements.
4. The students willing to opt OE under MOOC in their next semester, will submit their request to the MOOC Coordinator at their respective campus.
5. Once approved, the campus shall display the list of accepted courses.
6. The student has to submit certificate issued by the institution offering the MOOCs course, along with the number of credits and grades, to get credits transferred into his/her marks certificate issued by NDU.
7. The transfer of credits shall be as adopted by NDU.

Detailed Syllabus

Course Code	Course Name	L	T	P	Cr
DCSE250043	Cryptography and Network Security	3	0	2	4

Course Outcomes:

CO1: Explain security fundamentals, including the CIA Triad and various cyber threats.

CO2: Describe security services and mechanisms, such as authentication and encryption.

CO3: Understand the history and evolution of cryptography, including classical techniques.

CO4: Recognize the role of cryptanalysis in breaking cryptographic systems.

CO5: Apply mathematical foundations, including number theory, modular arithmetic, and random number generation, in cryptographic algorithms.

Module 1:

Introduction to Cryptography & Security Concepts

Basics of Security, Security Goals: Confidentiality, Integrity, Availability (CIA), Threats and Attacks: Passive vs. Active Attacks, Security Services & Mechanisms, Cybersecurity Principles, Introduction to Cryptography, History & Evolution of Cryptography, Classical Cryptographic Techniques: Caesar Cipher, Vigenère Cipher, Playfair Cipher, Modern Cryptography & Cryptanalysis

Mathematical Foundations of Cryptography: Number Theory Basics: Prime Numbers, Modular Arithmetic, Euler's Theorem, Greatest Common Divisor (GCD), Fermat's Theorem, Random Number Generation

Module 2:

Symmetric & Asymmetric Cryptography

Symmetric Key Cryptography, Block Ciphers vs. Stream Ciphers, Data Encryption Standard (DES) & Triple DES, Advanced Encryption Standard (AES), Modes of Operation: ECB, CBC, CFB, OFB, CTR, Asymmetric Key Cryptography, Public Key Cryptosystems: RSA Algorithm, Key Exchange Mechanisms: Diffie-Hellman Key Exchange, Elliptic Curve Cryptography (ECC), Comparison of Symmetric & Asymmetric Cryptography

Module 3:

Hash Functions, Digital Signatures & Authentication Cryptographic Hash Functions, Properties of Hash Functions, Common Hash Algorithms: MD5, SHA-1, SHA-256, SHA-3, Message Authentication, Message Authentication Codes (MAC), HMAC (Hash-based Message Authentication Code), Digital Signatures & Certificates, RSA Digital Signature Algorithm, Digital Certificates & Public Key Infrastructure (PKI), SSL/TLS Encryption and HTTPS Authentication Mechanisms, Password-Based Authentication, Multi-Factor Authentication (MFA), Biometric Authentication

Module 4:

Network Security & Secure Protocols

Network Security Fundamentals, Security in OSI & TCP/IP Models, Firewalls and Intrusion Detection Systems (IDS/IPS), Virtual Private Networks (VPNs) & IPsec, Secure Network Protocols, Secure Email: PGP & S/MIME, Secure Web Transactions: HTTPS, SSL/TLS

Wireless Security: WPA, WPA2, WPA3, Network Attacks & Defense Mechanisms, Denial of Service (DoS) & Distributed DoS (DDoS) Attacks, Man-in-the-Middle (MITM) & Eavesdropping, SQL Injection & Cross-Site Scripting (XSS)

Module 5:

Emerging Trends & Cybersecurity Applications

Blockchain & Cryptography, Role of Cryptography in Blockchain, Bitcoin & Ethereum Security, Cloud Security & Zero Trust Security Model, Cloud Encryption Mechanisms, Zero Trust Architecture & Access Control
Cybersecurity & Ethical Hacking, Introduction to Ethical Hacking, Penetration Testing Basics, Legal & Ethical Issues in Cybersecurity, Future Trends in Cryptography & Security

Labs / Practicals:

Preparing

References:

1. "The Joy of Cryptography" by Michael Rosulek
<https://joyofcryptography.com/pdf/book.pdf>
2. "Cryptography and Computer Security" by Chris Bourke
<https://cse.unl.edu/~cbourke/cryptoNotes.pdf>
3. "Applied Cryptography (20th Anniversary Edition)" by Bruce Schneier
<https://www.schneier.com/books/applied-cryptography>
4. "Cryptography and Network Security (Eighth Edition)" by William Stallings
<http://williamstallings.com/Cryptography/Crypto8e-Student>
5. "Cracking Codes With Python: An Introduction To Building And Breaking Ciphers" by Al Sweigart
<https://inventwithpython.com/cracking>
6. "Yet Another Introductory Number Theory Textbook (Cryptology Emphasis Version)" by Jonathan Adam Poritz
<https://www.poritz.net/jonathan/share/yaintt.pdf>

Course Code	Course Name	L	T	P	Cr
DASH250095	Research Methodology and IPR	3	1	0	4

Course Outcomes:

CO1: Understand the principles and process of research methodology.

CO2: Apply appropriate research design and data collection methods in computing projects.

CO3: Analyze and interpret qualitative and quantitative data using statistical tools.

CO4: Develop ethically sound, well-documented research or project reports conforming to academic and industrial standards.

CO5: Understand various forms of intellectual property and apply processes for protection and patent filing.

Module 1:

Introduction to Research and Research Ethics

Definition and objectives of research, types of research – fundamental, applied, exploratory, empirical, significance of research in computer applications, research ethics and integrity, plagiarism, citation styles (APA, IEEE), tools for plagiarism check.

Module 2:

Research Design and Data Collection

Research problem identification and formulation, hypothesis generation, literature review techniques, types of research design: experimental, descriptive, analytical, sampling techniques, primary and secondary data sources, survey design, case study design.

Module 3:

Data Analysis and Interpretation

Quantitative and qualitative data analysis, measures of central tendency and dispersion, t-test, ANOVA, correlation and regression, chi-square test, software tools: Excel, R, SPSS (demo-level), data visualization and result interpretation.

Module 4:

Technical Writing and Project Report Preparation

Structure of research papers and reports, abstract, introduction, literature review, methodology, results, and conclusions, referencing tools (Mendeley/Zotero), IEEE/ACM/SCI journal guidelines, thesis and dissertation formatting, proposal writing basics.

Module 5:

Intellectual Property Rights and Patent Process

Introduction to IPR: patents, copyrights, trademarks, industrial design, Indian and international patent systems (WIPO, TRIPS), criteria for patentability, patent filing process in India (IPINDIA portal), open-source licenses, patent drafting basics, IPR in software and AI innovations.

Labs / Practicals:

N/A

References:

1. C.R. Kothari & Gaurav Garg – Research Methodology: Methods and Techniques, New Age International.
2. Ranjit Kumar – Research Methodology – A Step-by-Step Guide, Sage Publications India.
3. P. Narayan – Intellectual Property Law, Eastern Book Company.
4. Neeraj Pandey & Khushdeep Dharni – Intellectual Property Rights, PHI Learning.
5. Yogesh Kumar Singh – Fundamentals of Research Methodology and Statistics, New Age International.
6. Bharat Bhushan – Research Methodology in Computer Science, Himalaya Publishing.
7. WIPO and IPIndia Manuals – Guidelines for Filing Patents in India.

Course Code	Course Name	L	T	P	Cr
DCSE250065	Digital Forensics & Cyber Crime Investigation	3	0	2	4

Course Outcomes:

CO1: Understand the languages of digital forensics, and the investigation of digital crime scene

CO2: Learn the basics of computer investigators

CO3: Become knowledgeable in the digital forensics networks and OSI layers

Module 1:

Introduction: Computer Forensics Needs, Computer forensics fundamentals, Introduction to Steps of Digital Forensics, Computer Crimes, Types of Digital forensics evidences, Legal Aspects of Digital Forensics.

Module 2:

Hardware and Software: Understanding Computer components- input and output devices, CPU, Digital Media, System software - Operating System Architecture, Application Software, File Systems, Memory organization concept, Data Storage concepts. Network: Topology, Devices, Protocols and Port, Communication media. IP Address: Types and classes.

Module 3:

Foundations: Basic Principles and methodologies for digital forensics, Design systems with forensic needs in mind. Phases of Digital Forensics. Introduction to Digital Forensics Tools, Life of a Digital Forensic Investigator. Data Acquisition: Principles of Digital Forensic Acquisition, Evidence Handling and Processing Digital Forensic Data.

Module 4:

Evidence Collection: Rules of Evidence, Jurisdictions, Techniques and standards for Preservation of Data. Evidence Analysis: OS /File System Forensics, Application Forensics, Web Forensics, Network Forensics, Mobile Device Forensics.

Module 5:

Investigation: Computer, Network, System attacks, Attack detection and investigation, Anti forensics. Case studies on File System, Network storage, Web and Mobile

Labs / Practicals:

NA

References:

1. Thomas J Holt , Adam M Bossler, Kathryn C Seigfried-Spellar, Cybercrime and Digital Forensics: An Introduction, Routledge, 2016
2. Eoghan Casey, Handbook of Digital Forensics and Investigation, Academic Press, 2017
3. Eoghan Casey, Digital Evidence and Computer Crime: Forensic Science, Computers, and the Internet, III Edition, 2016
4. Angus McKenzie Marshall, Digital Forensics: Digital Evidence in Criminal Investigations, Wiley- Blackwell, 2018

Course Code	Course Name	L	T	P	Cr
DCSA250021	Cyber Law	3	1	0	4

Course Outcomes:

CO1: Collect information using network scanning

CO2: Execute a penetration test using standard hacking tools in an ethical manner

CO3: Identify legal and ethical issues related to vulnerability and penetration testing

CO4: Plan a vulnerability assessment and penetration test for a network

CO5: Identify methods to gain access to systems

Module 1:

Emergence of Cyber space. Cyber Jurisprudence, Jurisprudence and law, Doctrinal approach, Consensual approach, Real Approach, Cyber Ethics, Cyber Jurisdiction, Hierarchy of courts, Civil and criminal jurisdictions, Cyberspace-Web space, Web hosting and web Development agreement, Legal and Technological Significance of domain Names, Internet as a tool for global access.

Module 2:

Overview of IT Act, 2000, Amendments and Limitations of IT Act, Digital Signatures, Cryptographic Algorithm, Public Cryptography, Private Cryptography, Electronic Governance, Legal Recognition of Electronic Records, Legal Recognition of Digital Signature Certifying Authorities, Cyber Crime and Offences, Network Service Providers Liability, Cyber Regulations Appellate Tribunal, Penalties and Adjudication.

Module 3:

Patent Law, Trademark Law, Copyright, Software – Copyright or Patented, Domain Names and Copyright disputes, Electronic Data Base and its Protection, IT Act and Civil Procedure Code, IT Act and Criminal Procedural Code, Relevant Sections of Indian Evidence Act, Relevant Sections of Bankers Book Evidence Act.

Module 4:

Relevant Sections of Indian Penal Code, Relevant Sections of Reserve Bank of India Act, Law Relating To Employees And Internet, Alternative Dispute Resolution , Online Dispute Resolution (ODR). Evolution and development in E-commerce, paper vs paper less contracts E-Commerce models- B2B, B2C, E security.

Module 5:

Application area: Business, taxation, electronic payments, supply chain, EDI, E-markets, Emerging Trends. Case Study On Cyber Crimes: Harassment Via E-Mails, Email Spoofing (Online A Method Of Sending E-Mail Using A False Name Or E-Mail Address To Make It Appear That The E-Mail Comes From Somebody Other Than The True Sender, Cyber Pornography (Exm.MMS),Cyber- Stalking

Labs / Practicals:

NA

References:

1. K. Kumar, "Cyber Laws: Intellectual property & ECommerce, Security", 1st Edition, Dominant Publisher, 2011.
2. Rodney D. Ryder, "Guide To Cyber Laws", Second Edition, Wadhwa And Company, New Delhi, 2007
3. Rodney D. Ryder, "Guide To Cyber Laws", Second Edition, Wadhwa And Company, New Delhi, 2007.

4. Information Security policy & implementation Issues, NIIT, PHI.
5. Vakul Sharma, "Handbook of Cyber Laws" Macmillan India Ltd, 2nd Edition, PHI, 2003.
6. Justice Yatindra Singh, "Cyber Laws", Universal Law Publishing, 1st Edition, New Delhi, 2003.
7. Sharma, S.R., "Dimensions Of Cyber Crime", Annual Publications Pvt. Ltd., 1st Edition, 2004.
8. Augustine, Paul T., "Cyber Crimes And Legal Issues", Crescent Publishing Corporation, 2007.

Course Code	Course Name	L	T	P	Cr
DCSE250092	Mobile & Digital Forensic	3	0	2	4

Course Outcomes:

CO1. Explain wireless technologies, identify security threats and vulnerabilities, and recommend protective measures against wireless and mobile attacks.

CO2. Analyze mobile communication security using the CIA triad and evaluate interception, codes, and SMS-related vulnerabilities in GSM systems.

CO3. Apply forensic techniques to extract, analyze, and preserve digital evidence from SIM cards, memory devices, and Android systems.

CO4. Demonstrate knowledge of digital forensics concepts, evidential potential of digital devices, and proper procedures for device handling and preservation.

CO5. Implement digital forensic examination principles, including imaging, hashing, audit logs, and evidence interpretation using structured security models.

Module 1:

Overview of wireless technologies and security: Personal Area Networks, Wireless Local Area Networks, Metropolitan Area Networks, Wide Area Networks. Wireless threats, vulnerabilities and security: Wireless LANs, War Driving, War Chalking, War Flying, Common Wi-Fi security recommendations, PDA Security, Cell Phones and Security, Wireless DoS attacks, GPS Jamming, Identity theft.

Module 2:

CIA triad in mobile phones-Voice, SMS and Identification data interception in GSM: Introduction, practical setup and tools, implementation- Software and Hardware Mobile phone tricks: Net monitor, GSM network service codes, mobile phone codes, catalog tricks and AT command set- SMS security issues.

Module 3:

Mobile phone forensics: crime and mobile phones, evidences, forensic procedures, files present in SIM card, device data, external memory dump, evidences in memory card, operators systems- Android forensics: Procedures for handling an android device, imaging android USB mass storage devices, logical and physical techniques.

Module 4:

Digital forensics: Introduction – Evidential potential of digital devices: closed vs. open systems, evaluating digital evidence potential- Device handling: seizure issues, device identification, networked devices and contamination.

Module 5:

Digital forensics examination principles: Previewing, imaging, continuity, hashing and evidence locations- Seven element security model- developmental model of digital systems- audit and logs- Evidence interpretation: Data content and context.

Labs / Practicals:

NA

References:

1. Gregory Kipper, "Wireless Crime and Forensic Investigation", Auerbach Publications, 2007
2. Iosif I. Androulidakis, "Mobile phone security and forensics: A practical approach", Springer publications, 2012
3. Andrew Hoog, "Android Forensics: Investigation, Analysis and Mobile Security for Google Android", Elsevier publications, 2011
4. Angus M. Marshall, "Digital forensics: Digital evidence in criminal investigation", John-Wiley and Sons, 2008

Course Code	Course Name	L	T	P	Cr
DOAI260001	AI (Industry)-Applied Ethics for AI	3	0	2	4

Course Outcomes:

CO1. Understand and be able to define and explain the terms 'AI for good' and 'responsible AI'. Describe the significance of responsible AI and recognize the potential benefits of AI in society.

CO2. Discuss and examine the impact of ethics on AI decision-making processes. Identify common ethical pitfalls in AI development and deployment. Explore methods for ethical AI practices.

CO3. Understand to apply an ethics-first approach to AI project planning and development. Compare different approaches to ethical AI design which include design of rules based on ethics and virtues. Identify factors influencing ethical decisions in AI development.

CO4. Apply Utilitarian and Kantian ethical principles to AI case studies, scenarios, and dilemmas. Explain the purpose and benefits of AI ethics codes. Develop a code of ethics for AI within an organization.

CO5. Identify the importance of interpretability in AI models. Evaluate the strengths and weaknesses of three different explainable AI tools: SHAP, LIME, and Tree Interpreter.

CO6. Conduct impact assessments for AI solutions. Apply ethical considerations in designing an AI system or solution. Implement measures to mitigate ethical risks and challenges post deployment.

Module 1:

Introduction to AI Ethics – meaning and importance, Role of Ethics in Responsible AI – defining 'AI for good' and 'responsible AI', significance of responsible AI in society, Common Ethical Pitfalls in AI – bias, fairness, accountability, transparency issues in AI development and deployment, Principles of Ethical AI – core principles guiding ethical AI, methods for ethical AI practices in development and deployment.

Module 2:

Ethics-First Approach in the AI Project Cycle – applying ethics at every stage of AI project planning and development, Approaches for Designing Ethical AI – rules-based ethics, values-based design, Introduction to Ethical Frameworks – overview of major ethical frameworks applied to AI, Virtue-Based Ethics – virtue ethics principles and the 4-box method for analysing ethical dilemmas, Applying Bio-Ethics Framework in AI – bioethical principles applied to AI in healthcare and biotechnology.

Module 3:

Utilitarianism for Ethical AI – utilitarian principles applied to AI case studies and dilemmas, Kantianism for Ethical AI – Kantian ethical principles applied to AI scenarios, Code of AI Ethics for Organizations – purpose, benefits and development of an organizational AI code of ethics, Data Protection and AI – risks of data breaches, privacy violations, responsible data handling.

Module 4:

Explainable AI – importance of interpretability in AI models, Intel XAI Toolkit, impact of explainability on decision-making and trust, Explainable AI in Practice – SHAP, LIME and Tree Interpreter tools: strengths, weaknesses and methods for applying them to analyze and interpret AI applications.

Module 5:

Impact Assessment of AI Solutions – conducting impact assessments for AI systems, Mitigating Ethical Issues Post Deployment – strategies to identify and resolve ethical risks after deployment, Project Creation – creating an AI startup with data card, model card, code of ethics, explainability report, impact self-assessment and transparency report.

Labs / Practicals:

1. Develop comprehensive Data Cards for datasets, as outlined in the Data Cards Playbook by Google, to ensure transparency and accountability in AI training data.
2. Design Responsible AI Best Practices.

3. Apply Human-Centered AI Canvas to design AI systems that address user needs and ethical considerations.
4. Learn to use The Box to embed AI principles into organizational workflows and operations.
5. Apply bio-ethics principles to AI case studies, scenarios, and dilemmas; evaluate the impact of virtue-based ethics on AI systems.
6. Develop AI ethics cards for given AI scenarios. Analyse a variety of ethical dilemmas by following the 4-box method from Virtue Ethics.
7. Analyse a variety of ethical dilemmas by walking through the ethical frameworks: Virtue Ethics, Bioethics, Utilitarian, Kantian, and Moral System Agnostic.
8. Develop an AI Code of Ethics for a fictitious company.
9. Use the UXAI Toolkit to brainstorm and plan different explainable interfaces for AI applications. Use the explainable AI library SHAP to create an explainability report for income prediction on the standard adult census income dataset.
10. Conduct impact assessments for AI solutions; create your own AI startup, develop business documents, write documentation (data card, model card, code of ethics, explainability report, impact self-assessment) and create and analyze transparency report.

References:

1. Intel AI for Future Workforce – Applied Ethics for AI Course Content, Intel Digital Readiness Program.
2. Virginia Eubanks – Automating Inequality: How High-Tech Tools Profile, Police, and Punish the Poor, St. Martin's Press.
3. Cathy O'Neil – Weapons of Math Destruction, Crown Publishers.
4. Kate Crawford – Atlas of AI: Power, Politics, and the Planetary Costs of Artificial Intelligence, Yale University Press.
5. Google Data Cards Playbook – <https://pair-code.github.io/datacardsplaybook/>

Course Code	Course Name	L	T	P	Cr
DOAI260002	AI (Industry)-Introduction to Artificial Intelligence	3	0	2	4

Course Outcomes:

- CO1. Describe what AI is and what it is not. Appreciate the history of AI and its evolution over time.
- CO2. Identify and analyze current trends in AI by correlating them with other technologies like IoT, Big Data and 5G.
- CO3. Identify 3 common domains of AI – Natural Language Processing, Computer Vision and Statistical Data – based on the type of underlying data.
- CO4. Examine and appreciate typical steps involved in an AI Project through the AI Project Cycle.
- CO5. Examine the immediate impacts of AI practices on society and appreciate the ethical concerns of AI applications.
- CO6. Classify Supervised Learning, Unsupervised Learning and Reinforcement Learning with the help of examples. Discuss the roles played by different key stakeholders in a typical AI team.

Module 1:

Demystification of AI – what AI is and what it is not, History and Evolution of AI, Current trends in AI, AI and Technology convergence, Industry 4.0 and Digitalization, Role of IoT, Big Data and 5G in AI, Domains of AI – Natural Language Processing, Computer Vision and Statistical Data.

Module 2:

AI Project Cycle-I – Problem Scoping and Data Acquisition, AI Project Cycle-II – Modelling and Evaluation, Societal Impact of AI – ethical concerns and responsible practices, Elements of ML – Supervised Learning, Unsupervised Learning and Reinforcement Learning with examples.

Module 3:

Elements of AI – types and characteristics, Data as Fuel for AI – structured and unstructured data, methods of data mining and data storage, AI/Data Science Team – roles and responsibilities of key stakeholders.

Module 4:

Tools to implement AI – No-Code tools (Teachable Machine, Orange, Roboflow, Chatfuel) and Code-based tools, Introduction to Neural Networks – working of simple Neural Networks, difference between common Deep Learning models with examples and applications.

Module 5:

AI Project-I – Natural Language Processing use case using no-code tool Chatfuel, AI Project-II – Statistical Data use case using no-code tool Orange Data Mining, AI Project-III – Computer Vision use case using no-code tool Roboflow, Future Possibilities of AI – emerging software and hardware technology trends and their direct impact on development of AI.

Labs / Practicals:

1. Familiarize with AI tools and techniques, including statistical data analysis using raw graphs, computer vision applications like Vision API, and generative AI tools such as DALL·E, Stable Diffusion, Gemini, and ChatGPT.
2. Apply the AI Project Cycle to real-world scenarios by identifying problems, collecting data, developing models, and evaluating their effectiveness to propose impactful AI-based solutions.
3. Deconstruct the working behind simple Neural Networks and outline the difference between some common DL models with the help of examples and applications.
4. Explore and utilize no-code tools, such as Teachable Machine, to build AI projects.
5. Project Building using No-code tool – Orange Data Mining for Statistical Data Use cases.
6. Project Building using No-code tool – Roboflow for Computer Vision Use cases.
7. Project Building using No-code tool – Chatfuel for Natural Language Processing Use cases.
8. Classify the type of data into structured and unstructured and evaluate data quality using real datasets.

9. Explore various AI domains by running Vision API experiments and comparing outcomes from NLP, CV, and Statistical Data use cases.
10. Demonstrate the AI Project Cycle on a chosen real-world problem from an industrial or social impact perspective and present findings.

References:

1. Intel AI for Future Workforce – Introduction to AI Course Content, Intel Digital Readiness Program.
2. Stuart Russell and Peter Norvig – Artificial Intelligence: A Modern Approach, Pearson Education.
3. Andreas Mueller and Sarah Guido – Introduction to Machine Learning with Python, O'Reilly Media.
4. Ian Goodfellow, Yoshua Bengio, and Aaron Courville – Deep Learning, MIT Press.
5. Aurélien Géron – Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow, O'Reilly Media.

Course Code	Course Name	L	T	P	Cr
DOAI260003	AI (Industry)-Introduction to Generative AI	3	0	2	4

Course Outcomes:

CO1. Define Gen AI and distinguish its role in the broader context of AI. Describe how the basic model of Gen AI works. Discuss the wider societal impact of Gen AI.

CO2. Understand how NLP has evolved over time and the shift from Traditional AI to Gen AI. Understand Prompt Engineering and its impact on the output of NLP tasks.

CO3. Learn the limitations of current AI generation tools for video and audio. Recognize the challenges in ethical considerations and responsibilities when using generative AI tools.

CO4. Describe the stages of a generative AI project and learn to select and use appropriate AI tools for development.

CO5. Understand the basic structure of neural networks, architecture of transformer models, workings of diffusion models, and the pre-training process for generative AI models.

CO6. Learn to fine-tune generative AI models, understand RLHF, prototype Gen AI applications, and apply principles of responsible and ethical use of LLMs.

Module 1:

Introduction to Gen AI – definition and scope, role of Gen AI in the broader context of AI, how the basic model of Gen AI works, societal impact of Gen AI, Demystifying Gen AI Models – types of generative models (GANs, VAEs, Diffusion Models, LLMs), evolution of NLP to Gen AI, shift from Traditional AI to Gen AI.

Module 2:

Prompt Engineering-I – introduction to prompt engineering, basic applications and impact on NLP task outputs, text generation and conversation initiation, Prompt Engineering-II – advanced prompt engineering techniques for diverse NLP applications, real-world examples such as Mint Mobile Commercial, Basics of Video and Audio for Generation using Gen AI for Productivity-I – introduction to video generation tools (Runway ML, Genmo), text-to-video and image-to-video generation, limitations of current AI generation tools for video and audio.

Module 3:

Basics of Video and Audio for Generation using Gen AI for Productivity-II – audio and music generation using AIVA, code generation using Codeium, text generation using ChatGPT, Gemini and Perplexity, Generative AI-Ethics – ethical considerations, responsibilities and challenges when using generative AI tools, Generative AI Project Lifecycle – stages of a Gen AI project, selection of appropriate AI tools, Gen AI Tools for Development – open-source tools such as Stable Diffusion XL, proprietary tools such as Bing Image Creator powered by DALLE3.

Module 4:

Neural Network Fundamentals – basic structure and functioning of neural networks, how neural networks process data to make predictions, Transformers and the Evaluation of LLMs – architecture of transformer models, methods to evaluate large language models for performance and reliability, walkthrough of OpenAI Playground and parameter exploration, Diffusion Models and the Evaluation of Text to Image Generation – workings of diffusion models, evaluation of quality of text-to-image generation, Pre-Training Gen AI Models-I – pre-training process for generative AI models, importance of pre-training in building effective AI systems, modifying classification labels of Cohere datasets and its impact on fine-tuning and model confidence.

Module 5:

Fine-Tuning Gen AI Models – techniques for fine-tuning generative AI models to adapt them for specific tasks and applications, RLHF – Evaluating and Optimizing Gen AI Models: reinforcement learning with human feedback, evaluation and improvement of performance of generative AI models, Prototyping Gen AI Applications – skills to design, develop and test prototypes of generative AI applications, LangChain components – LLM, Agents, Prompts, Chains, Document loaders and utils with code-based implementation, LLM Application and Responsible use of AI – practical applications of large language models, ethical and responsible use of LLMs, building an AI

Tutor using open-source models such as Dolly 3B and Intel Neural Chat 7B.

Labs / Practicals:

1. Apply prompt engineering skills in basic NLP tasks, such as text generation, conversation initiation, and simple problem-solving.
2. Develop advanced prompt engineering skills for diverse NLP applications. Examine real-world prompt engineering examples, such as the Mint Mobile Commercial, to discern its role in marketing.
3. Explore cutting-edge Gen AI tools, their features, and capabilities for content generation including video and audio.
4. Explore open-source models such as Stable Diffusion XL and proprietary tools such as Bing Image Creator (powered by DALL-E3) for image generation.
5. Text-to-video generation and image-to-video generation using tools like Runway ML and Genmo. Generate audio and music files using the tool AIVA.
6. Generate code by prompts using the tool Codeium. Generate text using tools like ChatGPT, Gemini, and Perplexity.
7. Walkthrough of OpenAI Playground. Explore how changing parameters can affect responses generated by different models.
8. Explore how modifying classification labels of Cohere datasets impacts fine-tuning and model confidence.
9. Explore basic components of LangChain like LLM, Agents, Prompts, Chains, and Document loaders and utils, along with a code-based implementation.
10. Utilize open-source text generation models such as Dolly 3B and Intel Neural Chat 7B for building an AI Tutor application.

References:

1. Intel AI for Future Workforce – Introduction to Generative AI Course Content, Intel Digital Readiness Program.
2. Antony Hagiwara – Generative Deep Learning: Teaching Machines to Paint, Write, Compose, and Play, O'Reilly Media.
3. Ashish Vaswani et al. – Attention is All You Need, Advances in Neural Information Processing Systems.
4. OpenAI – GPT-4 Technical Report, OpenAI.
5. Harrison Chase – LangChain Documentation – <https://docs.langchain.com/>

Course Code	Course Name	L	T	P	Cr
DOAI260004	AI (Industry)-Introduction to Machine Learning	3	0	2	4

Course Outcomes:

- CO1. Describe Machine Learning and how Deep Learning is a subset of the wider field of Machine Learning.
- CO2. Understand the importance of dashboards and discuss various applications of dashboards across different industries.
- CO3. Explain the mathematical working behind different ML models. Implement different classification and regression ML models using Python.
- CO4. Outline the difference between supervised, unsupervised and reinforcement learning algorithms. Examine the working of different reinforcement learning models.
- CO5. Describe the working of Neural Networks and contrast it with biological Neurons. Describe various applications of neural networks. Outline different methods to overcome variance and bias in DL models.
- CO6. Discuss and interpret the future of ML based on current and upcoming trends. Develop Python-based AI Projects incorporating Supervised Learning, Unsupervised Learning and Deep Neural Networks.

Module 1:

Introduction to Machine Learning – definition, types and applications, Relationship between AI, ML and Deep Learning, Tools to Implement AI (Python & Mathematics) – I: Python basics for ML, NumPy, Pandas, data manipulation.

Module 2:

Tools to Implement AI (Python & Mathematics) – II: mathematical foundations – linear algebra, statistics and probability, No-Code tool for Data Exploration – Tableau dashboard creation, data visualization techniques, statistical concepts implementation using no-code visualization tools.

Module 3:

AI (Modeling) Supervised Learning-I – Linear Regression, Logistic Regression, mathematical working behind models, classification and regression using Python, AI (Modeling) Supervised Learning-II – SVM, Decision Tree, industrial applications of ML models, AI (Modeling) Unsupervised Learning-I – K-Means Clustering, mathematics of clustering algorithms, AI (Modeling) Unsupervised Learning-II – Recommendation Systems, mathematics behind recommendation system, applications of recommendation systems.

Module 4:

Reinforcement Learning – definition, difference from supervised and unsupervised learning, working of reinforcement learning models with applications, Neural Network and Deep Learning-I – biological vs artificial Neurons, working of Neural Networks, applications of neural networks, Deep Learning-II – Convolutional Neural Networks, Recurrent Neural Networks, methods to overcome variance and bias in DL models, implementation using Python and deep learning frameworks, exploring OpenAI's Gym for reinforcement learning.

Module 5:

ML Project – Supervised Learning AI models: end-to-end Python project using classification/regression, ML Project – Unsupervised Learning AI models: end-to-end Python project using clustering, ML Project – Deep Neural Networks: building and training DNN in Python, Future of Machine Learning and Deep Learning – current and upcoming trends in ML and DL.

Labs / Practicals:

1. Interpret and understand the basic tools required for building ML Projects through a selected list of topics from Mathematics and Python Programming.
2. Develop and create a simple dashboard for visualizing data through Tableau.
3. Implement statistical concepts using the no-code visualization tool and create your own dashboard with the help of a no-code visualization tool.

4. Implement supervised machine learning algorithms, such as SVM and Decision Tree, using Python to develop a deeper understanding of their practical applications.
5. Implement unsupervised machine learning algorithms, such as K-Means, using Python to develop a deeper understanding of their practical applications.
6. Implement neural networks and create simple generative AI models using Python and deep learning frameworks, including exploring OpenAI's Gym for hands-on experience with AI and reinforcement learning.
7. Develop Python-based use cases and AI Projects incorporating Supervised Learning methods.
8. Develop Python-based use cases and AI Projects incorporating Unsupervised Learning methods.
9. Develop Python-based use cases and AI Projects incorporating Deep Neural Networks.
10. Implement a recommendation system using Python and evaluate its performance on a real-world dataset.

References:

1. Intel AI for Future Workforce – Introduction to Machine Learning Course Content, Intel Digital Readiness Program.
2. Andreas Mueller and Sarah Guido – Introduction to Machine Learning with Python, O'Reilly Media.
3. Ian Goodfellow, Yoshua Bengio, and Aaron Courville – Deep Learning, MIT Press.
4. Aurélien Géron – Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow, O'Reilly Media.
5. Christopher Bishop – Pattern Recognition and Machine Learning, Springer.

Course Code	Course Name	L	T	P	Cr
DCSE250019	Cloud Architecture and Security	3	0	2	4

Course Outcomes:

CO1: Explain cloud models (Public, Private, Hybrid) and services (IaaS, PaaS, SaaS) with their benefits and challenges.

CO2: Describe cloud application deployment and compare platforms like AWS, Azure, and Google Cloud.

CO3: Apply basic cloud security concepts like authentication, access control, and data protection.

CO4: Identify virtualization and multi-tenancy issues, including VM isolation and recovery.

CO5: Outline cloud security management standards and data security practices across service models.

Module 1:

Computing Fundamental: Cloud computing definition, private, public and hybrid cloud. Cloud aS, PaaS, SaaS. Benefits and challenges of cloud computing, public vs private clouds, role of ation in enabling the cloud; Business Agility: Benefits and challenges to Cloud architecture.

Module 2:

Cloud Applications: Technologies and the processes required when deploying web services-Deploying a web service from inside and outside a cloud architecture, advantages and disadvantages- Development environments for service development; Amazon, Azure, Google App.

Module 3:

Security Concepts: Confidentiality, privacy, integrity, authentication, nonrepudiation, availability, access control, defence in depth, least privilege- how these concepts apply in the cloud and their importance in PaaS, IaaS and SaaS. e.g. User authentication in the cloud.

Module 4:

Multi-tenancy Issues: Isolation of users/VMs from each other- How the cloud provider can provide this- Virtualization System Security Issues: e.g. ESX and ESXi Security, ESX file system security- storage considerations, backup and recovery- Virtualization System Vulnerabilities.

Module 5:

Management in the cloud – security management standards- SaaS, PaaS, IaaS availability ent- access control- Data security and storage in cloud and recovery- Virtualization System Vulnerabilities

Labs / Practicals:

NA

References:

1. GautamShroff,Enterprise Cloud Computing Technology Architecture Applications [ISBN: 978-0521137355]
2. Toby Velte, Anthony Velte, Robert Elsenpeter, Cloud Computing, A Practical Approach [ISBN: 0071626948]
3. Tim Mather, SubraKumaraswamy, ShahedLatif, Cloud Security and Privacy: An Enterprise Perspective on Risks and Compliance [ISBN: 0596802765]
4. Ronald L. Krutz, Russell Dean Vines, Cloud Security [ISBN: 0470589876]

Course Code	Course Name	L	T	P	Cr
DCSE250093	Mobile & Wireless Security	3	0	2	4

Course Outcomes:

CO1: Describe wireless technologies and their security vulnerabilities.

CO2: Identify mobile network threats and attack types.

CO3: Explain physical layer security techniques in wireless systems.

CO4: Analyze security challenges in ad hoc wireless networks.

CO5: Evaluate RFID security and privacy protection methods.

Module 1:

WIRELESS NETWORK SECURITY THREATS AND VULNERABILITIES

Introduction to wireless technologies, Design Factors, security threats and vulnerabilities present at the different protocol layers, family of security protocols and algorithms used in the existing wireless networks (Bluetooth, Wi-Fi, Wi MAX and LTE standards).

Module 2:

Introduction To Mobile Network Techs, Vulnerabilities Threats And Attack Entry Points. Categorization Of Attacks In Mobile Networks, Signaling Attacks. Threats And Attacks In 4g Networks- Attacks Against Security And Confidentiality, Ip-Based Attacks, Gtp-Based Attacks, Volte Sip-Based Attacks, Diameter-Based Attacks.

Module 3:

Emerging physical layer security in wireless communications. Class of informationTheoretic security, artificial-noise-aided security, security-oriented beam forming, security- oriented diversity, and physical-layer secret key generation techniques. Review on various wireless jammers, open challenges in wireless security.

Module 4:

Security in Ad Hoc Wireless Networks, Network Security Requirements, Issues and Challenges in Security Provisioning, Network Security Attacks, Key Management in Adhoc Wireless Networks, Secure Routing in Adhoc Wireless Networks.

Module 5:

Introduction, RFID Security and privacy, RFID chips Techniques and Protocols, RFID anti- counterfeiting, Man-in-the-middle attacks on RFID systems, Digital Signature Transponder, Combining Physics and Cryptography to Enhance Privacy in RFID Systems, Scalability Issues in Large-Scale Applications, An Efficient and Secure RFID Security Method with Ownership Transfer, Policy-based Dynamic Privacy Protection Framework leveraging Globally Mobile RFIDs, User-Centric Security for RFID based Distributed Systems, Optimizing RFID protocols for Low Information Leakage, RFID: an anti-counterfeiting tool.

Labs / Practicals:

NA

References:

1. Kia Makki, Peter Reiher, "Mobile and Wireless Network Security and Privacy ", Springer, ISBN 978-0-387-71057-0, 2007.
2. Siva Ram Murthy.C, Manoj B.S, "Adhoc Wireless Networks Architectures and By Yulong Zou, Senior Member IEEE, Jia Zhu, Xianbin Wang, Senior Member IEEE, and Lajos Hanzo, Fellow IEEE
3. "A Survey on Wireless Security: Technical Challenges, Recent Advances, and Future Trends" Zou et al.:

A Survey on Wireless Security: Technical Challenges, Recent Advances, and Future Trends

4. Nouredine Boudriga, "Security of Mobile Communications",
5. ISBN 9780849379413, 2010.
6. Kitsos, Paris; Zhang, Yan, "RFID Security Techniques, Protocols and System-On-Chip Design", ISBN 978-0-387-76481-8, 2008.
7. Johnny Cache, Joshua Wright and Vincent Liu," Hacking Wireless Exposed: Wireless Security Secrets & Solutions ", second edition, McGraw Hill, ISBN: 978-0-07-166662-6, 2010.

Course Code	Course Name	L	T	P	Cr
DOAI250017	Data Analytics for Fraud Detection	3	0	2	4

Course Outcomes:

CO1: Understand various types of fraud, their characteristics, and the role of data analytics in fraud detection.

CO2: Apply data preprocessing and exploratory data analysis techniques on real-world fraud datasets.

CO3: Build and evaluate machine learning models for detecting fraudulent activities.

CO4: Implement unsupervised and real-time fraud detection techniques using big data tools.

CO5: Analyze ethical, legal, and privacy considerations in fraud analytics and present practical fraud detection solutions.

Module 1:

Introduction to Fraud and Data Analytics :

What is Fraud? Types of Fraud (Financial, Identity Theft, Insurance, Healthcare, Cyber Fraud, etc.), Fraud Triangle Theory and Red Flags, Importance and Role of Data Analytics in Fraud Detection, Overview of the Fraud Detection Lifecycle, Data Analytics Workflow in Fraud Detection, Case Studies: Enron, Wells Fargo, etc.

Module 2:

Data Preprocessing and Exploration for Fraud Analytics :

Data Collection Methods and Challenges, Data Cleaning and Normalization Techniques, Handling Imbalanced Data (Oversampling, SMOTE, Undersampling), Feature Engineering: Domain-Specific Feature Creation, Exploratory Data Analysis (EDA): Detecting Anomalies and Patterns, Visualization Techniques for Fraud Trends

Module 3:

Machine Learning Techniques for Fraud Detection :

Supervised Learning Methods: Logistic Regression, Decision Trees, Random Forest, XGBoost, Unsupervised Learning: Clustering, Isolation Forest, Autoencoders, One-Class SVM, Evaluation Metrics: Precision, Recall, F1-Score, ROC-AUC, Model Validation and Cross-Validation Techniques, Model Interpretability: SHAP, LIME

Module 4:

Real-Time and Advanced Fraud Detection Systems :

Introduction to Streaming Data and Real-Time Fraud Detection, Anomaly Detection in Streaming Data using Spark, Kafka, Rule-based vs AI/ML-based Detection Systems, Big Data Platforms for Fraud Analytics: Hadoop, Spark, Introduction to Graph-Based Fraud Detection (Social Networks, Transaction Networks)

Module 5:

Ethics, Case Studies, and Industry Applications :

Ethics in Fraud Analytics: Privacy, Fairness, Bias, Legal Aspects and Compliance (AML, GDPR, PCI-DSS), Fraud Detection in Banking, Insurance, Healthcare, E-Commerce, Recent Trends: AI in Fraud Detection, Blockchain Use

Labs / Practicals:

1. Perform EDA on a credit card fraud dataset to identify trends and anomalies.
2. Clean and preprocess transactional data by handling missing values and normalization.
3. Apply SMOTE or undersampling to balance an imbalanced fraud detection dataset.
4. Build and evaluate a logistic regression model for fraud classification.
5. Compare Decision Tree and Random Forest models on fraud detection accuracy.
6. Implement Isolation Forest or KMeans for unsupervised fraud detection.
7. Simulate real-time fraud alerts using Kafka and Spark streaming.
8. Use SHAP or LIME to interpret a fraud detection model's predictions.
9. Model transaction data as a graph and detect fraud rings using NetworkX.
10. Create an end-to-end fraud detection project from data cleaning to model deployment.

References:

1. "Fraud Analytics: Strategies and Methods for Detection and Prevention" – Bart Baesens
2. "Data Science for Business" – Foster Provost and Tom Fawcett
3. "Practical Fraud Prevention" – Gilit Saporta
4. Research papers, white papers, and industry reports from SAS, KPMG, and Deloitte

Course Code	Course Name	L	T	P	Cr
DCSA250035	Information Security Audit	3	0	2	4

Course Outcomes:

CO1: Discussed various algorithms and Distributions.

CO2: Understanding the approaches of message authentication

CO3: Analyze the security principles and its requirements

CO4: Apply the roles and procedures for audit

CO5: Analyze the approaches to audits during the system development

Module 1:

A model for Internetwork security, Conventional Encryption Principles & Algorithms (DES, AES, RC4, Blowfish), Block Cipher Modes of Operation, Location of Encryption Devices, Key Distribution. Public key cryptography principles, public key cryptography algorithms (RSA, Diffie- Hellman, ECC), public Key Distribution.

Module 2:

Approaches of Message Authentication - Secure Hash Functions (SHA-512, MD5) and HMAC, Digital Signatures, Kerberos, X.509 Directory Authentication Service, Email Security: Pretty Good Privacy (PGP) IP Security: Overview, IP Security Architecture, Authentication Header, Encapsulating Security Payload, Combining Security Associations and Key Management.

Module 3:

Web Security: Requirements, Secure Socket Layer (SSL) and Transport Layer Security (TLS), Secure Electronic Transaction (SET). Firewalls: Firewall Design principles, Trusted Systems, Intrusion Detection Systems.

Module 4:

Auditing For Security: Introduction, Basic Terms Related to Audits, Security audits, The Need for Security Audits in Organization, Organizational Roles and Responsibilities for Security Audit, Auditors Responsibility In Security Audits, Types Of Security Audits.

Module 5:

Information Security Assessments: Vulnerability Assessment, Classification, Types of Vulnerability Assessment, Vulnerability Assessment Phases, Vulnerability Analysis Stages, Characteristics of a Good Vulnerability Assessment Solutions & Considerations, Vulnerability Assessment Reports – Tools and choosing a right Tool, Information Security Risk Assessment, Risk Treatment, Residual Risk, Risk Acceptance, Risk Management Feedback Loops etc.

Labs / Practicals:

NA

References:

1. Information Security by Mark Stamp, Wiley-INDIA, 2006.
2. Fundamentals of Computer Security, Springer.
3. Network Security: The complete reference, Robert Bragg, Mark Rhodes, TMH
4. Computer Security Basics by Rick Lehtinen, Deborah Russell & G. T. Gangemi Sr., SPD O'REILLY 2006.
5. Modern Cryptography by Wenbo Mao, Pearson Education 2007.
6. Principles of Information Security, Whitman, Thomson.

Course Code	Course Name	L	T	P	Cr
DCSE250049	Data Privacy	3	0	2	4

Course Outcomes:

CO1: To introduce the fundamentals of statistics, data privacy & policies.

CO2: To Study the mathematical model and computing practices

CO3: To learn the protection models and surveys

CO4: To study the computation system

CO5: Aware of policies and practices of technology

Module 1:

Data Privacy and its Importance- Need for Sharing Data, Methods of Protecting Data, Importance of Balancing Data Privacy and Utility, Disclosure, Tabular Data, Micro data, Approaches to Statistical disclosure control, Ethics, principles, guidelines and regulations.

Module 2:

Micro data-Disclosure, Disclosure risk, Estimating re-identification risk, Non-perturbative micro data masking, Perturbative micro data masking, Information loss in micro data.

Module 3:

Static Data Anonymization on Multidimensional Data -Privacy Preserving Methods, Classification of Data in a Multidimensional Data Set, Group- Based Anonymization, k-Anonymity, l-Diversity, t-closeness.

Module 4:

Static Data Anonymization on Complex Data Structures -Privacy Preserving Graph Data, Privacy Preserving Time Series Data, Time Series Data Protection Methods, Privacy Preservation of Longitudinal Data, Privacy Preservation of Trans- action Data.

Module 5:

Data Anonymization Threats-Threats to Anonymized Data, Threats to Data Structures, Threats by Anonymization Techniques, Randomization, k- Anonymization, l-Diversity, tCloseness. Dynamic Data Protection: Tokenization, Understanding Tokenization, Use Cases for Dynamic Data Protection, Benefits of Tokenization Compared to Other Methods, Components for Tokenization.

Labs / Practicals:

NA

References:

1. George T. Duncan. Mark Elliot, Juan-Jose Salazar-Gonzalez, Statistical Confidentiality: Principle and Practice. Springer, 2011. (ISBN No.: 978-1-44-197801-1).
2. Aggarwal, Charu C., Yu, Philip S., Privacy-Preserving Data Mining: Models and Algorithms, Springer, 2010. (ISBN No.: 978-0-38-770991-8). Mode of Evaluation: CAT / Assignment / Quiz / FAT / Project / Seminar

Course Code	Course Name	L	T	P	Cr
DCSE250072	Edge Computing	3	1	0	4

Course Outcomes:

- CO1: This course will explore research, frameworks, and applications in Edge Computing
 CO2: The class will begin with are view of current IoT Applications
 CO3: Explore frameworks for computing using Raspberry Pi
 CO4: Apply the Interfacing edge to cloud applications
 CO5: Analyze edge computing with others

Module 1:

Introduction to Edge Computing Scenario's and Use cases - Edge computing purpose and definition, Edge computing use cases, Edge computing hardware architectures, Edge platforms, Edge vs Fog Computing, Communication Models - Edge, Fog and M2M.

Module 2:

A connected ecosystem, IoT versus machine-to-machine versus, SCADA, The value of a network and Metcalfe's and Beckstrom's laws, IoT and edge architecture, Role of an architect, Understanding Implementations with examples - Example use case and deployment.

Module 3:

Introduction to Raspberry Pi, About the Raspberry Pi Board: Hardware Layout and Pinouts, Operating Systems on Raspberry Pi, Configuring Raspberry Pi, Programming Raspberry Pi, Connecting Raspberry Pi via SSH, Remote access tools, Interfacing DHT Sensor with Pi, Pi as Web server, Pi Camera, Image & Video Processing using Pi.

Module 4:

Implementation of Microcomputer Raspberry Pi and device Interfacing, Edge to Cloud Protocols, MQTT, MQTT publish-subscribe, MQTT architecture details, MQTT state transitions, MQTT packet structure, MQTT data types, MQTT communication formats, MQTT 3.1.1 working example.

Module 5:

Edge computing with Raspberry Pi, Industrial and Commercial IoT and Edge, Edge computing and solutions, Case study – Telemedicine palliative care, Requirements, Implementation, Use case retrospective.

Labs / Practicals:

NA

References:

1. Fog and Edge Computing: Principles and Paradigms by Rajkumar Buyya, Satish Narayana Srirama, Wiley publication, 2019, ISBN: 9781119524984.
2. David Jensen, "Beginning Azure IoT Edge Computing: Extending the Cloud to the Intelligent Edge, MICROSOFT AZURE

Course Code	Course Name	L	T	P	Cr
DCSA250005	Applied Cryptography	3	1	0	4

Course Outcomes:

- CO1: Understand the fundamentals of number theory and algorithms.
 CO2: Analyze, design and implement different cryptography protocols.
 CO3: Apply the intermediate protocols for linking and distributing.
 CO4: Understand various Security practices and System security standards.
 CO5: Apply the various Authentication schemes to simulate different applications.

Module 1:

Number theory: Fermat's and Euler's theorem-Chinese remainder theorem-Euclidean algorithm- Test for primality-Discrete logarithms, Information theory: entropy, Uncertainty Complexity theory: pseudo random number generation and generators.

Module 2:

Protocol Building Blocks-Basic Protocols: key Exchange-Authentication-Authentication and Key exchange: Wide-mouth frog, Yahalom, Kerberos-Formal Analysis of Authentication and Key Exchange Protocols-Multiple Key Public Key Cryptography-Secret Splitting-Secret Sharing: Secret Sharing with Cheaters-Cryptographic protection of Databases.

Module 3:

Time stamping services, Linking protocol, Distributed Protocol-Udeniable digital signatures Proxy Signatures- Group Signatures-Fail-stop signatures-computing with encrypting data-bit commitment- Fair coin flips-one-way accumulators.

Module 4:

Zero knowledge proof, Parallel Zero Knowledge Proof, Zero Knowledge proof of identity: Chess Grandmaster Problem-Blind Signatures-Simultaneous Contract Signing-Digital certified Mail- Simultaneous Exchange of Secrets-Esoteric protocols: Secure Elections-Secure Multiparty Computation.- Digital cash.

Module 5:

Key Length: Symmetric key Length, Public Key length-Algorithm types and Modes: Electronic Code Book Mode, Block Replay, Cipher Block Chaining Mode-Using Algorithms: Choosing an Algorithm, Public Key Cryptography vs Symmetric Cryptography, Encrypting Communication Channels- Public Key Algorithms: RSA, Pohlig-Hellman, Rabin, Elliptic Curve Cryptosystems - Public Key Digital Signature Algorithms: Ghost Digital Signature Algorithm, Discrete Logarithm Signature schemes. Real World approach: IBM secret key management protocol- MITRENET, ISDN, SESAME.

Labs / Practicals:

NA

References:

1. Applied Cryptography: Protocols, Algorithms and source code in C, Wiley, Second Edition- Bruce Schneier (Oct 18, 1996)
2. Cryptography and Network Security Principles and practices- William Stallings (Jan 24, 2010)
3. Foundations of Cryptography: Volume 1, Basic Tools by Oded Goldreich (Jan 18, 2007)
4. Encryption: High-impact Strategies- What You Need to Know: Definitions, Adoptions, Impact, Benefits, Maturity... by Kevin Roebuck, Emerepty Limited, 2011.
5. Foundations of Cryptography: Volume 2, Basic Applications by Oded Goldreich (Sep 17, 2009)

Course Code	Course Name	L	T	P	Cr
DOAI250032	Machine Learning	3	0	2	4

Course Outcomes:

CO1: Identify and extract relevant features from raw IoT data suitable for applying appropriate Machine Learning (ML) techniques.

CO2: Evaluate and compare the strengths and limitations of various ML algorithms across diverse application domains, particularly in IoT.

CO3: Select and implement appropriate ML models based on data characteristics and application requirements.

CO4: Mathematically formulate and analyze ML models and algorithms, including their underlying assumptions and performance metrics.

CO5: Design and validate machine learning-based solutions for real-world problems in IoT, demonstrating practical understanding and technical proficiency.

Module 1:

Module 1

Introduction: Introduction to Machine Learning: Introduction. Different types of learning, Hypothesis space and inductive bias, Evaluation. Training and test sets, cross validation, Concept of over fitting, under fitting, Bias and Variance.

Linear Regression: Introduction, Linear regression, Simple and Multiple Linear regression, Polynomial regression, evaluating regression fit.

Module 2:

Module 2

Decision tree learning: Introduction, Decision tree representation, appropriate problems for decision tree learning, the basic decision tree algorithm, hypothesis space search in decision tree learning, inductive bias in decision tree learning, issues in decision tree learning, Python exercise on Decision Tree.

Instance based Learning: K nearest neighbor, the Curse of Dimensionality, Feature Selection: forward search, backward search, univariate, multivariate feature selection approach, Feature reduction (Principal Component Analysis), Python exercise on KNN and PCA. Recommender System: Content based system, Collaborative filtering based.

Module 3:

Probability and Bayes Learning: Bayesian Learning, Naïve Bayes, Python exercise on Naïve Bayes, Logistic Regression.

Support Vector Machine: Introduction, the Dual formulation, Maximum margin with noise, nonlinear SVM and Kernel function, solution to dual problem.

Module 4:

Artificial Neural Networks: Introduction, Biological motivation, ANN representation, appropriate problem for ANN learning, Perceptron, multilayer networks and the back propagation algorithm.

Module 5:

Ensembles: Introduction, Bagging and boosting, Random forest, Discussion on some research papers.

Clustering: Introduction, K-mean clustering, agglomerative hierarchical clustering, Python exercise on k-mean clustering.

Labs / Practicals:

1 Write a Python program that imports a dataset of your choice, displays the first five rows, and prints the data type of the DataFrame as a whole. Then, generate and display the statistical summary (mean, median, standard deviation, etc.) for each column and provide complete information about the DataFrame.

- 2 Write a Python program to import (or create) a dataset with missing values, display the count of missing values per column, and handle them using both imputation and dropping methods. Then, compare the dataset before and after handling missing values, and display the final cleaned dataset.
- 3 Write a Python program to load (or create) a dataset, split it into training and testing sets, and standardize the features using StandardScaler from sklearn. preprocessing. Finally, display the dataset before and after standardization to observe the changes.
- 4 Write a Python program to load (or create) a dataset, separate it into predictor variables (X) and target variable (y), and then split the dataset into training and testing sets. Finally, print both the training and testing datasets.
- 5 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 Python ML library classes in the program.
- 6 Write a Python program to load (or create) a dataset with missing values, count missing values in each column, and handle them by replacing with NaN, the column mean, and the most frequent value. Finally, display the dataset before and after performing the replacements.
- 7 Write a Python program to load (or create) a dataset, split it into training and testing sets, and apply feature scaling to the data. Finally, display the dataset before and after scaling to observe the changes.
- 8 Write a program to implement k-Nearest Neighbour algorithm to classify the iris data set. Print both correct and wrong predictions. Python ML library classes can be used for this problem.
- 9 Implement the working principle of Artificial Neural Network with feed forward and feed backward principle.
- 10 Implement and demonstrate the working of SVM algorithm for classification.

References:

1. Kevin P. Murphy, "Machine Learning: A Probabilistic Perspective", MIT Press, 2012.
2. C. M. Bishop. Pattern Recognition and Machine Learning. First Edition. Springer, 2006. (Second Indian Reprint, 2015).
3. Tom M. Mitchell, "Machine Learning", McGraw-Hill, 2010
4. P. Flach. Machine Learning: The Art and Science of Algorithms that Make Sense of Data. First Edition, Cambridge University Press, 2012.

Course Code	Course Name	L	T	P	Cr
DOEE250124	IOT and its Applications	3	1	0	4

Course Outcomes:

CO1: Understand the basics of IoT.

CO2: Implement the state of the Architecture of an IoT.

CO3: Understand design methodology and hardware platforms involved in IoT.

CO4: Understand how to analyze and organize the data.

CO5: Compare IOT Applications in Industrial & realworld.

Module 1:

FUNDAMENTALS OF IoT- Evolution of Internet of Things, Enabling Technologies, M2M Communication, IoT World Forum (IoTWF) standardized architecture, Simplified IoT Architecture, Core IoT Functional Stack, Fog, Edge and Cloud in IoT, Functional blocks of an IoT ecosystem, Sensors, Actuators, Smart Objects and Connecting Smart Objects.

Module 2:

IoT PROTOCOLS- IoT Access Technologies: Physical and MAC layers, topology and Security of IEEE 802.15.4, 802.11ah and Lora WAN, Network Layer: IP versions, Constrained Nodes and Constrained Networks, 6LoWPAN, Application Transport Methods: SCADA, Application Layer Protocols: CoAP and MQTT.

Module 3:

DESIGN AND DEVELOPMENT- Design Methodology, Embedded computing logic, Microcontroller, System on Chips, IoT system building blocks
IoT Platform overview: Overview of IoT supported Hardware platforms such as: Raspberry pi, Arduino Board details

Module 4:

DATA ANALYTICS AND SUPPORTING SERVICES:

Data Analytics: Introduction, Structured Versus Unstructured Data, Data in Motion versus Data at Rest, IoT Data Analytics Challenges, Data Acquiring, Organizing in IoT/M2M,

Supporting Services: Computing Using a Cloud Platform for IoT/M2M

Applications/Services, Everything as a service and Cloud Service Models.

Module 5:

CASE STUDIES/INDUSTRIAL APPLICATIONS: IoT applications in home, infrastructures, buildings, security, Industries, Home appliances, other IoT electronic equipments, Industry 4.0 concepts.

Labs / Practicals:

NA

References:

1. The Internet of Things – Key applications and Protocols, Olivier Hersent, David Boswarthick, Omar Elloumi and Wiley, 2012 (for Unit2).
2. "From Machine-to-Machine to the Internet of Things – Introduction to a New Age of Intelligence", Jan Ho"ller, Vlasios Tsiatsis, Catherine Mulligan, Stamatis,

Karnouskos, Stefan Avesand. David Boyle and Elsevier, 2014.

3. Architecting the Internet of Things, Dieter Uckelmann, Mark Harrison, Michahelles and Florian (Eds), Springer, 2011.

4. Recipes to Begin, Expand, and Enhance Your Projects, 2nd Edition, Michael Margolis, Arduino Cookbook and O'Reilly Media, 2011.

Course Code	Course Name	L	T	P	Cr
DCSA250008	Biometrics	3	1	0	4

Course Outcomes:

CO1: Understand biometric traits, image processing concepts, and error analysis for biometric system design.

CO2: Apply various filtering and edge detection techniques to enhance and analyze biometric images.

CO3: Explain biometric system components, authentication methods, and feature extraction using PCA.

CO4: Evaluate system performance using FAR/FRR, ROC/DET curves, and assess security vs. usability trade-offs.

CO5: Analyze multi-modal biometric systems, identify vulnerabilities, and apply recognition techniques for fingerprints, faces, signatures, ears, and irises.

Module 1:

Introduction of biometric traits and its aim. Image processing, pattern recognition, statistics, error types. What is image, acquisition, types. Point operations, geometric transformations. Linear interpolation, brightness correction, histogram.

Module 2:

Basic image operations: convolution, linear/non-linear filtering. Gaussian, median, min filters, gray level reduction. Special filters, enhancement filters. Edge detection, derivatives, Laplacian, unsharp masking. High-boost filtering, sharpening, edge detection steps. First and second derivatives, smoothing, thresholding, localization. Robert's, Sobel's, Prewitt methods, Laplacian of Gaussian, Zero crossing. Canny edge detection. Fourier series, DFT, inverse DFT.

Module 3:

Biometric system, authentication. Physiological and behavioral traits, biometric system properties. Application areas of biometrics. PCA, eigenvectors/values, 2D-PCA, generalization to p-dimensions. Covariance, correlation, PCA algebra, data projection.

Module 4:

Identification vs. verification, threshold, score distribution. FAR/FRR, system design issues. Positive/negative identification, authentication methods and protocols. Hypothesis testing (H_0 , H_1), error types I/II. Matching score distribution, FM/FNM, ROC, DET, FAR/FRR curves. Comparing systems using ROC, EER, biometric myths. Trade-offs: security vs. convenience. Biometric selection, Zephyr charts, multi-biometric types.

Module 5:

Multi-model system verification, normalization, fusion methods. Multi-modal identification, biometric system security. System vulnerabilities: circumvention, covert acquisition. Quality control, template generation, data storage, interoperability. Signature recognition system: cropping, enhancement, matching. DWT, face detection, feature template, matching. Fingerprint recognition: enhancement, thinning, minutiae, CN number. Ear and iris recognition: image acquisition, cropping, normalization.

Labs / Practicals:

NA

References:

1. Digital Image Processing using MATLAB, By: Rafael C. Gonzalez, Richard Eugene Woods, 2nd Edition, Tata McGraw-Hill Education 2010
2. Guide to Biometrics, By: Ruud M. Bolle, Sharath Pankanti, Nalini K. Ratha, Andrew W. Senior, Jonathan H. Connell, Springer 2009
3. Pattern Classification, By: Richard O. Duda, David G. Stork, Peter E. Hart, Wiley 2007

Course Code	Course Name	L	T	P	Cr
DCSA250009	Blockchain Technology	3	1	0	4

Course Outcomes:

Upon completion of the course, the learners will be able to:

CO1: Understand the concept of blockchain and its need.

CO2: Analyze methods of cryptography for application with blockchain.

CO3: Evaluate the working of blockchain.

CO4: Understand the underlying technology of transactions, blocks, proof-of-work, and consensus building.

CO5: Identify real world problems that blockchain can solve and analyze a use case.

Module 1:

Fundamentals of Blockchains Technology, The Structure of Blockchains, Blockchain Applications, The Blockchain Life Cycle, Distributed Ledger vs Centralized System, Merkel Root, Nonce Value, Hash Algorithm, and its applications.

Module 2:

Block Ciphers; Encryptions; Secret Keys; Elliptic Curve Cryptography; Hash cryptography; Encryption vs hashing; Digital Signature; Memory Hard Algorithm, Zero Knowledge Proof.

Module 3:

Introduction: Transactions, blocks, hashes, consensus, verify and confirm blocks, peer to peer networks, blocks of data in a chain, decentralization of networks, processes & workflows, cryptocurrencies, nodes, assets, consensus; types of blockchain; chain policy; working of blockchain; life of blockchain application; privacy, anonymity and security of blockchain.

Module 4:

Hyperledger: Introduction, where can Hyperledger be used, Hyperledger architecture, Hyperledger Fabric, features of Hyperledger; Open source blockchain platform technology; Tools & services; Cloud options in blockchains AWS; Azure workbench; Hyperledger console; Consensus: Proof of work, proof of stake, delegated proof of stake, proof of burn, BlocBox Protocol.

Module 5:

History; Distributed ledger; Smart contracts; Cryptocurrency; Bitcoin protocols; Mining strategy and rewards; Ethereum – construction; DAO; GHOST; Vulnerability; Attacks; Sidechain; Namecoin

Labs / Practicals:

NA

References:

1. D. Tapscott, A. Tapscott, Blockchain Revolution: How the Technology Behind Bitcoin Is Changing Money, Business, and the World, Portfolio Publishers.
2. A. Antonopoulos, Mastering Bitcoin: Programming the Open Blockchain, O'Reilly.
3. P. Champagne, The Book of Satoshi: The Collected Writings of Bitcoin Creator Satoshi Nakamoto, e53 Publishing, LLC.
4. M. Swan, Blockchain: Blueprint for a New Economy, O'Reilly.

5. R. Wattenhofer, The Science of the Blockchain, Inverted Forest Publishing.
6. R. Modi, Solidity Programming Essentials: A beginner's guide to build smart contracts for Ethereum and blockchain, Pack

Course Code	Course Name	L	T	P	Cr
DCSA250033	Image Forensics and Security	3	0	2	4

Course Outcomes:

- CO1: Upon successful completion of this course, the students will get an in-depth knowledge in image and video forensics and its security techniques
- CO2: Helps students to learn various types of image formation Techniques
- CO3: Students will learn the Fourier Transform and Forensic image analysis
- CO4: Students will gain the knowledge of statistical based forensics.
- CO5: Students learn to conduct a Video Forensics & Image Security Techniques in an organized and systematic way.

Module 1:

Introduction to Image Processing, Background, Digital Image Representation, Fundamental steps in Image Processing, Elements of Digital Image Processing- Image Acquisition, Storage, Processing, Communication, Display.

Module 2:

Image formation, image compression, point processing, neighbourhood operations, image analysis. Morphological Image Processing : Dilation and Erosion, Opening and Closing, Extensions to gray level images, hit or miss transformation, basic morphologic algorithms.

Module 3:

Format-Based Forensics- Fourier Transform-Smoothing and Sharpening, frequency domain filters- Ideal, Butterworth and Gaussian Filters, Homomorphic filtering, JPEG, Camera-Based Forensics. Pixel-Based Forensics: Resampling, Cloning, Thumbnails.

Module 4:

Principal Component Analysis, Linear Discriminant Analysis, Quadratic Discriminant Analysis and Logistic Regression, Computer Generated or Photographic: Perception.

Module 5:

Motion, Re-Projected, Projectile, Enhancement Physics-Based Forensics: 2-D Lighting, Lee Harvey Oswald (case study). Image Hiding, Image Coding. Image file Forensics, Video Surveillance, RFID and Vehicular tracking (GPS) devices, Image security techniques: visual cryptography, Stenography, water marking.

Labs / Practicals:

NA

References:

1. N. Efford, Digital Image Processing, Addison Wesley 2000, ISBN 0-201-59623-7
2. M Sonka, V Hlavac and R Boyle, Image Processing, Analysis and Machine Vision, PWS 1999, ISBN 0-534-95393
3. Pratt. W.K., Digital Image Processing, John Wiley and Sons, New York, 1978