Institute Level- Industrial orientated Open Elective OPEN ELECTIVE OTHER THAN PARTICULAR PROGRAM (OE) AIDSML

Structure

| | | SEMESTER | Ш | | | | | | |
|------------|----------------|--|---|---|---|---------|-----|-----|-------|
| Sr. No. | Course Code | Course | L | Т | P | Credits | ISE | ESE | Total |
| 1. | IOE3315 | Open Elective I-Foundations of AI, Data Science, and Machine Learning | 3 | | | 3 | 50 | 50 | 100 |
| 2. | IOE3316 | Open Elective-01 Lab - Foundations of AI, Data Science, and Machine Learning Lab | | | 2 | 1 | 25 | 25 | 50 |
| | | SEMESTER I | V | | | | | | |
| 3. | IOE3417 | Open Elective II - AI in Practice: Design to Deployment | | | | 2 | 50 | 50 | 100 |
| | | SEMESTER | V | | | | | | |
| 4. | IOE3514 | Open Elective III- AI Applications and Emerging Technologies | 2 | | | 2 | 50 | 50 | 100 |
| | | TOTAL | 7 | | 2 | 08 | 175 | 175 | 350 |

Note * The End Semester Examination (ESE) will be conducted as either theory or oral or presentation

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Rajdrillen.

BoS Chairman
Dept. of Electronics and Telecommunication

Institute Level- Industrial orientated Open Elective OPEN ELECTIVE OTHER THAN PARTICULAR PROGRAM (OE) AIDSML

| | Second | Year (Sem - | III) OE- Institute Lev | el- Industrial ori | entated Open El | lective- A | IDSM | L | |
|--|---|--|--|--|---------------------|------------------|--|----------|--|
| | | | en Elective I-Foundatio | | | | | | |
| Teach | ing Sche | eme | | ns of fri, Data Sel | Examination Scho | omo | g | | |
| Lectur | | 03 Hrs/week | | | ISE | 50 | | | |
| Tutori | als | 00 Hrs/week | | | ESE | 50 | ACCOUNT OF THE PARTY OF THE PAR | | |
| Total (| Credits | 03 | | | Duration of ESE | As appl | icable | | |
| | | | | | | | | icabic | |
| Prere | quisite : | Mathematics, St | atistics and Programming | for problem solvir | ng | | | | |
| Cours | e Outco | mes: Students w | vill be able to | The second control of | 8 | 1 - 1/1-1/15/20 | Carlot C | | |
| CO1 | Understand the fundamental principles and concepts of Artificial Intelligence (AI) and Data Sci | | | | | | | | |
| CO2 | Demo (ML) | onstrate a solid u | inderstanding of the math | ematical foundation | ns essential for AI | and Mach | ine Lear | ning | |
| CO3 | Apply | y Python program | mming skills to perform of | lata manipulation a | nd analysis | | | | |
| CO4 | Empl | oy techniques fo | ortime series analysis and | demonstrate in den | th understanding o | f Machine | Learni | nσ | |
| | | | Course (| Contents | | 1 Iviaciiiiic | CO | Hour | |
| | Intro | duction to AI & | & Data Science: | | | | | Hour | |
| Unit 1 | Over | view of AI and | Data Science, The dat | a science workflo | w. AI application | s across | CO1 | (05) | |
| | vario | us industries, Et | hical considerations in AI | and data science | , apparention | 3 461 655 | COI | (00) | |
| | Math | ematical and S | tatistical Foundations fo | or AI & ML: | | | | | |
| Unit 2 | | Linear algebra basics: vectors, matrices, and operations, Calculus essentials: derivatives and | | | | | | | |
| | integr | als, Probability | and statistics for data scie | ence. | | | | (07) | |
| | Prog | Programming Fundamentals for AI & Data Science | | | | | | | |
| Unit 3 | Pytho | Python for data manipulation, Libraries: NumPy and Pandas for data science, Data | | | | | | | |
| | to To | visualization with Matplotlib and Seaborn, Introduction to Scikit-learn for AI. Introduction to TensorFlow and PyTorch. | | | | | | | |
| | Dete | Propressessing | Torch. | | | | | | |
| Unit 4 | Hand | Data Preprocessing Techniques and Exploratory Data Analysis (EDA): | | | | | | | |
| UIIII 4 | Handling Missing Data, Data Cleaning and Noise Removal, Data Transformation, Encoding | | | | | | | (06) | |
| | and T | and Feature Engineering, Data exploration and visualization techniques | | | | | | | |
| | Time | Time Series Analysis and Data Science Life Cycle: | | | | | | 8 ha = 1 | |
| Unit 5 | nine | Time series components (trend, seasonality, noise), Moving averages, smoothing techniques, autocorrelation. Data Science Life Cycle and Overview of MLOps. | | | | | | | |
| | autoc | orrelation. Data | Science Life Cycle and O | verview of MLOps | 5. | | | (07) | |
| | Fund | amentals of Ma | chine Learning: | | | | | | |
| | Overv | Overview and implementation basics of various machine learning. Supervised Learning: | | | | | | | |
| Unit 6 | Definition, examples, and algorithms; linear regression, decision trees. SVM Unsupervised | | | | | | CO4 | (07) | |
| | Learn | Learning: Definition, examples, and algorithms k-means clustering, hierarchical clustering | | | | | | (07) | |
| | PCA. | | | | | , | | | |
| Text B | | | | | | | | | |
| 1. $\begin{vmatrix} w \\ 20 \end{vmatrix}$ | es McK: | inney, "Python i | for Data Analysis: Data V | Wrangling with Par | idas, NumPy, and | Python" (|)'Reilly | Media | |
| 20 | 11. | | | | | | | | |
| 2. A | oplication | ns in R" Springe | tten, Trevor Hastie, and | Robert Hibshirani - | Introduction to S | Statistical | Learnin | g: with | |
| 3 Sa | njeev J. | Wagh, Manish | a S. Bhende, Anuradha I | Thakare "Funda | mentals of Data C | olomos T- | 10 | г : | |
| | cc press | 2021. | | | | | yler & | Fransic | |
| 4 A | an Beau | lieu - "Learning | SQL: Generate, Manipula | ate, and Retrieve D | ata" - O'Reilly Med | dia 2009 | | | |
| keiere | nce Bool | KS | | | | | | | |
| 1 1 | oel Grus | - "Data Science | from Scratch: First Princ | iples with Python" | - O'Reilly Media 2 | 2015 | | | |
| 1. J | ırélien (19. | Géron - "Hands- | On Machine Learning v | vith Scikit-Learn, I | Keras, and Tensor | Flow" - (| D'Reilly | Media | |
| \rightarrow | | | | | | | | | |
| 2. Al | Links | | | Control of the Contro | | | net de la constant | | |
| 2. All 20 Iseful | | necourses.nptel. | ac.in/noc21 cs69/preview | 1 | | E | 1 22 | | |
| 2. 20 Useful 1. htt | ps://onli | necourses.nptel. | ac.in/noc21_cs69/preview ac.in/noc22_cs32/preview | 1 | | <u> </u> | Janes - | | |

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Govt. College of Engineering, KARAD

| | | Governmen | t College of Engineering, Karad | | | | |
|---|--|--|---|---------|--|--|--|
| Seco | ond Year | | nstitute Level- Industrial orientated Open Elect | ive- | | | |
| IOE221 | | EL OLT L E | AIDSML | | | | |
| | | | undations of AI, Data Science, and Machine Learni | ng Lab | | | |
| | tory Sche | | Examination Scheme: | Mess . | | | |
| Practica T-t-LC | | 02 Hrs/week | ISE 25 | | | | |
| Total Cr | | 01 | ESE 25 | | | | |
| | | | nd Programming for problem solving | 1117 | | | |
| Course | | es (CO):Students will l | | | | | |
| CO1 | CO1 Understand the fundamental principles of data science, AI applications, and scripting. | | | | | | |
| CO2 | Apply P | ython programming sk | tills to perform data manipulation, analysis, and visuali | zation | | | |
| CO3 | Demons | trate proficiency in lin | ear algebraic computations and exploratory data Analy | sis | | | |
| CO4 | | | essing techniques and Time Series Analysis | | | | |
| | 100000000000000000000000000000000000000 | | ourse Contents | CO | | | |
| Implem | entation | of following concepts | S | Likest | | | |
| Experin | nent 1 | Data Science Work a sample dataset. | flow: Implement a basic data science workflow using | CO1 | | | |
| Experiment 2 | | AI Applications : Case study analysis of AI applications in healthcare, finance, and retail. | | | | | |
| Experin | nent 3 | Python Basics : Write Python scripts for basic data operations (CRUD - Create, Read, Update, Delete). | | | | | |
| Experin | nent 4 | NumPy: Perform array operations and linear algebraic computations using NumPy. | | | | | |
| Experin | nent 5 | | Pandas: Data manipulation and analysis using Pandas (e.g., merging, grouping, and aggregating data). | | | | |
| Experiment 6 Matplotlib and Seaborn: Create various types of plots (line, bar, scatter) using Matplotlib and Seaborn. | | | | | | | |
| Experiment 7 Handling Missing Values: Techniques to handle missing data (e.g., imputation, deletion). | | | | | | | |
| Experiment 8 Linear Algebra: Implement matrix operations, eigenvalues, and eigenvectors using Python. | | | | | | | |
| Experin | nent 9 | End to End Data Wrangling: Scaling, Feature Engineering, Outlier Handing etc. | | | | | |
| Experin | nent 10 | Exploratory Data Analysis (EDA): Perform EDA on a dataset to summarize its main characteristics. | | | | | |
| Experin | nent 11 | Visualization: Creat data distributions. | te histograms, box plots, and pair plots to visualize | CO4 | | | |
| Experin | nent 12 | Forecasting using b | asic Time Series Analysis | CO4 | | | |
| List of S | ubmissio | | • | | | | |
| | | | Minimum number of Experim | ents: 1 | | | |

*Note: End Sem Exam (ESE) will be conducted either practical/oral or presentation mode.

Govt. College of Engineering, KANAR

| S | econd Year (Sem – IV) OE- Institu | t College of Engineering, Kar- te Level- Industrial orientated | | - AIDS | ML | | |
|---------------------------|--|---|-------------------|--------------|---------|--|--|
| | | ve II - AI in Practice: Design to | | V 1112 51 | | | |
| Teachi | ng Scheme | Examination Sch | | | | | |
| Lecture | | ISE | 50 | 1 | | | |
| Tutoria | | ESE | 50 | | | | |
| Total Credits | 02 | Duration of ESE | As applicable | | | | |
| | | | | 1 200 | | | |
| | uisite: Foundations of AI, Data Science | | | A GO | | | |
| | Outcomes (CO): Students will be able t | | | | | | |
| CO1 | Implement supervised and unsupervi | | The Mag. | Blob R 3 | | | |
| CO2 | Enhance model performance through | | lection. | | | | |
| CO3 | Develop and apply CNNs and RNNs | | | | | | |
| CO4 | Utilize advanced data mining techniq | | orks for Compute | | | | |
| | | ourse Contents | -, | CO | Hours | | |
| Unit 1 | Introduction to Scikit-learn library, Implementation of algorithms like lin and SVM using Scikit-learn, | Supervised Machine Learning with Python: Introduction to Scikit-learn library, Implementing Supervised Learning Algorithms: Implementation of algorithms like linear regression, logistic regression, decision trees, and SVM using Scikit-learn. | | | | | |
| Unit 2 | Implementing Unsupervised Learnin | Unsupervised Machine Learning with Python: Implementing Unsupervised Learning Algorithms: Implementation of algorithms like k-means clustering, hierarchical clustering using Scikit-learn. | | | | | |
| Unit 3 | performance., Model Selection: Strategies for selecting the best model, cross-validation, and hyperparameter tuning. | | | | | | |
| Unit 4 | Neural Networks (CNNs), Recurrent Neural Networks (RNNs): Structure, applications, and implementation basics | | | | (06) | | |
| Unit 5 | Natural Language Processing (NLI Text processing: tokenization, st vectorization methods and word emb | CO3 | (04) | | | | |
| Unit 6 | Computer Vision and Advanced Da Computer Vision Fundamentals: Ima recognition, | CO4 | (03) | | | | |
| Text Bo | | | | | | | |
| 1. Eth | em Alpaydin - "Introduction to Machine | Learning" - MIT Press (2020) | | | | | |
| 2. Au (20 | rélien Géron - "Hands-On Machine Le | arning with Scikit-Learn, Keras, | and TensorFlow' | - O'Reil | lly Med | | |
| | hard Szeliski - "Computer Vision: Algor | ithms and Applications" - Springe | r (2010) | | | | |
| Referer | ice Books | | | | | | |
| (20 | | | | | | | |
| Dat | ce Zheng and Amanda Casari - "Featur a Scientists" - O'Reilly Media (2018) | | | | | | |
| 3. S. J pres Useful | J. Wagh, Manisha S. Bhende, Anuradha ss 2021 | a D. Thakare "Fundamentals of D | ata Science, Tayl | er & Fra | nsic CR | | |
| | | | | The state of | | | |
| | tps://nptel.ac.in/courses/106102220/ tps://nptel.ac.in/courses/106106145/ | | | 1 | | | |
| | tps://nptel.ac.in/courses/106106212/ | 7 | | 1 | | | |
| | -po iiptoi.ac.ii/ oouises/ 100100212/ | | and l | mela | OL A | | |

*Note: End Sem Exam (ESE) will be conducted either theory or oral or presentation mode.

Jepartment of Information Technology Govt. College of Engineering, KALLED

| 7 | Third Va | ar (Sam V | | College of Engineer Level- Industrial or | | otivo A | IDCMI | |
|--------------------------------------|--|---------------------------------|-----------------------|--|-------------------------------------|---------------|----------------------|-------------|
| | | | | AI Applications an | | | | 1 |
| Caachin | | OE3314: Op | en Elective III- 2 | At Applications an | Examination Sch | | | |
| Teaching Scheme Lectures 02 Hrs/week | | | | ISE | 50 | CONTRACT OF | | |
| Futorials | | Hrs/week | | | ESE | 50 | | |
| Total Cre | | | | | Duration of ESE | | nlicable | |
| otal Cit | zuits 02 | | | | Duration of ESE | As ap | plicable | |
| Prorogu | icito: Adv | anced AI Integ | gration | | | | MINE TO THE STATE OF | |
| | | | ts will be able to | | | | Table 1 | |
| CO1 | | | | ent and explore advar | and tooksisses like | | al CANIa | |
| | Utilize v | ector database | es for semantic ret | rieval and Analyze a | and use Large Langu | onumon Mor | dolo (LLI | (1a) fo |
| CO2 | | enerative task | | rievai aliu Alialyze a | illu use Laige Laiigu | lage Mo | dels (LLI | VIS) 10 |
| | | | | output with external | 1 knowledge and A | nly pro | mnt engi | naarin |
| CO3 | | | cific responses | output with externa | r knowledge and Ap | ppry pro | inpt engi | neer m |
| CO4 | | | | nms and apply them in | n autonomous system | 10 | 149 | |
| CO5 | Denloy A | I on edge dev | ices and integrate | with IoT for application | one in smart cities in | duetry a | nd health | care |
| 000 | Deploy 1 | on eage dev | | rse Contents | Jus in smart cities, in | dustry, a | CO | Hou |
| | Generat | ve Adversari | | Ns) and Creative AI: | | | CO | 1100 |
| | | | | nd their architecture, | | Me in | | |
| Unit 1 | generatin | g realistic im | ages, videos, and cr | reative content, Explo | oring conditional GA | Ns and | CO1 | (04 |
| | style tran | sfer technique | es. Case studies in a | art, design, and conter | it creation | 145 dild | | |
| | style transfer techniques, Case studies in art, design, and content creation. Introduction to Generative AI: Fundamentals of Generative AI, Vector Databases, | | | | | | | |
| Unit 2 | | | | use cases: content g | | | CO2 | (04 |
| | image sy | | | | | , | | (. |
| | Generative AI: Retrieval-Augmented Generation (RAG), Prompt Engineering, Fine- | | | | | | | |
| Unit 3 | Tuning of LLMs, LoRA (Low-Rank Adaptation), PEFT (Parameter-Efficient Fine- | | | | | | | |
| Unit 3 | Tuning). Hugging Face Transformers. Real-world applications: chatbots, search assistants, | | | | | | | (05 |
| | enterprise AI | | | | | | | |
| | Reinforcement Learning and Autonomous Systems: | | | | | | | |
| Unit 4 | Introduction to reinforcement learning principles, Applications of reinforcement learning | | | | | | | (0.4) |
| Unit 4 | in autonomous systems, Deep dive into algorithms such as Q-learning and deep Q- | | | | | | | (04 |
| | networks, Case studies on robotics, gaming, and control systems. | | | | | | | |
| | Edge AI and Internet of Things (IoT) Integration: | | | | | | | |
| Unit 5 | Deploying AI algorithms on edge devices for real-time processing, Integration of AI with | | | | | | | (05 |
| Omit 5 | IoT ecos | ystems for si | mart applications, | Use cases in smart | cities, industrial Io | T, and | CO4 | (05) |
| | healthcar | e monitoring, | Challenges and opp | portunities in edge AI | and IoT convergenc | e. | | |
| | | | d Biomedical Appl | | | | | |
| Unit 6 | Role of AI in medical imaging analysis and diagnosis, AI-driven drug discovery and | | | | | | | (0.4) |
| | personalized medicine, Patient care management using AI-based solutions, Ethical | | | | | | CO5 | (04) |
| | considera | tions and regu | alatory challenges i | n AI-driven healthcar | e. | | | |
| ext Boo | | | | | | | | |
| Dav | id Foster | "Generative | Deep Learning: T | eaching Machines to | Paint, Write, Comp | ose, and | Play" - | O'Reil |
| Med | 11a (2019) | | | | | | | |
| 2. Max | Cim Lapan | - "Deep Rein | forcement Learning | g Hands-On" - Packt 1 | Publishing (2018) | | | |
| 3. Bna | rat S. Cha | udhari, Sheet | al N. Ghorpade, M | arco Zennaro, "Tinyl | ML for Edge Intellig | ence in | IoT and I | LPWA |
| | | | mic Press (2024). | | | | | |
| l. Perr | e Books | and Edge Co | imputing for Archit | tects - Second Edition | Paperback – Import | , 6 Marcl | n 2020 | |
| | | "Overture M | (auliina I aurii N | 71 + 0 | | | | |
| (201 | 6) | Quantum M | achine Learning: w | Vhat Quantum Compu | iting Means to Data I | Mining" | - Academ | nic Pre |
| S K | | II Havit Gree | ensnan Dinggang | Shen - "Deep Learni | ing for Modical La- | go A = -1 | raia!! A | 201- |
| Pres | s (2017) | a, mayn Gree | mspan, Dinggang | onen - Deep Learn | ing for Medical Ima | ge Analy | sis" - A | cadem |
| | | and Daniel Si | tunavake - "TinvM | L: Machine Learning | with TensorFlow I | ite on A | rduine en | d I II+- |
| Low | -Power N | licrocontroller | rs" - O'Reilly Media | a (2020) | , with Tellsoffiow L | D a | | |
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| | | ac.in/courses/ | 106106139/ | | | EAD | 2 | |
| l. htt | . F1. | ac.m/com/scs | | | | | | |
| | | | | N. Control of the Con | | TEAD | on Toon | noine |
| | | ac.in/courses/ | | | Separtment of It Ggyt: Gollage o | formanti | on Toor | inclos |