

# A Six Months Course

on

## IMPLEMENTATION OF AI, ML AND DL ALGORITHMS USING PYTHON

### DETAILED COURSE STRUCTURE



---

#### COURSE

#### COORDINATOR:

**Dr. B. SATEESH KUMAR**

*Professor & Head,*

*Department of CSE,*

**JNTUH UNIVERSITY COLLEGE OF ENGINEERING JAGTIAL.**

## **STRUCTURE OF AI COURSE**

**SUBJECT NAME**

### **1. ARTIFICIAL INTELLIGENCE**

<b>UNIT-1</b>	<b>INTRODUCTION TO AI</b>
<b>DAY-1</b>	Definition and Scope of AI History and Evolution of AI AI Agent and Environment Examples and Types of AI: Narrow vs. General AI vs. Super AI
<b>DAY-2</b>	AI Techniques Logic concepts and Logic Programming Knowledge Representation and Inference Rule-Based Systems
<b>UNIT-2</b>	<b>EXPERT SYSTEMS</b>
<b>DAY-3</b>	Introduction Phases in Building Expert Systems Expert System Architecture Expert Systems Vs. Traditional Systems
<b>DAY-4</b>	Principles of Expert Systems Truth Maintenance Systems Expert System Shells and Tools Case Studies: Expert Systems in Healthcare, Finance, and Engineering
<b>UNIT-3</b>	<b>NATURAL LANGUAGE PROCESSING (NLP)</b>
<b>DAY-5</b>	Fundamentals of NLP: Tokenization, Part-of-Speech Tagging, and Parsing Named Entity Recognition (NER) and Sentiment Analysis
<b>DAY-6</b>	Sequence-to-Sequence Models: Encoder-Decoder Architectures, Language Modeling: Introduction, N-Gram Models, Language Model Evaluation, Parameter Estimation, Language Model Adaptation, Types of Language Models, Language-Specific Modeling Problems,
<b>DAY-7</b>	Multilingual and Cross Lingual Language Modeling , Advanced NLP Applications: Machine Translation, Text Summarization, and Question Answering Systems
<b>UNIT-4</b>	<b>KNOWLEDGE REPRESENTATION AND REASONING</b>
<b>DAY-8</b>	Knowledge Representation Techniques: Case Grammars, Semantic Networks, Frames, and Ontologies
<b>DAY-9</b>	Logic-Based Reasoning: Propositional and First-Order Logic, Uncertainty and Probabilistic Reasoning
<b>DAY-10</b>	Case Studies: Knowledge-Based Systems in Decision Support and Planning
<b>UNIT-5</b>	<b>ROBOTIC PROCESS AUTOMATION (RPA) AND GENERATIVE AI</b>
<b>DAY-11</b>	Introduction to RPA, RPA Tools and Technologies: Overview of popular RPA platforms (e.g., UiPath, Automation Anywhere, Blue Prism)

	<b>DAY-12</b> Understanding the components of RPA software (bots, orchestrators, control room), Selection criteria for choosing an RPA tool
	<b>DAY-13</b> Steps involved in RPA implementation (identification, assessment, design, development, testing, deployment), Use Cases of RPA
	<b>DAY-14</b> Introduction to Generative AI, Overview of different generative models (GANs, VAEs, autoregressive models), Generative AI in Practice Real-world applications of Generative AI (image synthesis, text-to-image generation, style transfer, etc.)
<b>UNIT-6</b>	<b>AI APPLICATIONS</b>
	<b>DAY-15</b> Linguistic aspects of natural language processing, A.I. And Quantum Computing, Applications of Artificial Intelligence (AI) in business
	<b>DAY-16</b> Emotion Recognition using human face and body language, AI based system to predict the diseases early, Smart Investment analysis
	<b>DAY-17</b> AI in Sales and Customer Support, Robotic Processes Automation for supply chain management, AI-Optimized Hardware
	<b>DAY-18</b> Digital Twin i.e. AI Modelling, Information Technology & Security using AI, AI/ML in Smart solutions, AI/ML in Social Problems handling, Block chain and AI
<b>PRACTICAL SESSION-1</b>	
	<b>DAY-19</b> Developing and testing rule-based expert systems in Python
	<b>DAY-20</b> Developing logic-based and probabilistic reasoning algorithms in Python
	<b>DAY-21</b> Implementing NLP tasks using NLTK/spaCy
<b>PRACTICAL SESSION-2</b>	
	<b>DAY-22</b> Implementing RPA tasks using Python libraries
	<b>DAY-23</b> Building and training generative models for image synthesis or text generation tasks
	<b>DAY-24</b> Hands-on project: Developing an end-to-end AI application using Python

## STRUCTURE OF ML COURSE

**SUBJECT NAME**

### **2. MACHINE LEARNING**

<b>UNIT-1</b>	<b>FOUNDATIONS OF MACHINE LEARNING</b>
<b>DAY-1</b>	Basics of Supervised, Unsupervised, Semi-supervised and Reinforcement Learning Types of Machine Learning Algorithms: Classification, Regression, Clustering, and Reinforcement Learning
<b>DAY-2</b>	Overview of the machine learning pipeline: data preprocessing, model training, evaluation, and deployment
<b>UNIT-2</b>	<b>SUPERVISED LEARNING TECHNIQUES</b>
<b>DAY-3</b>	Linear Regression and Logistic Regression Decision Trees and Random Forests
<b>DAY-4</b>	Support Vector Machines (SVMs), K-Nearest Neighbors (k-NN), Naive Bayes Classifier, Bayesian Neural Networks
<b>DAY-5</b>	Elastic Net Regression, Ordinal Regression, Gaussian Processes, Ensemble Methods: Bagging and Boosting
<b>UNIT-3</b>	<b>UNSUPERVISED LEARNING TECHNIQUES</b>
<b>DAY-6</b>	K-Means Clustering, Hierarchical Clustering, DBSCAN (Density-Based Spatial Clustering of Applications with Noise), Mean Shift Clustering
<b>DAY-7</b>	Principal Component Analysis (PCA), Association Rule Learning: Apriori Algorithm
<b>DAY-8</b>	Gaussian Mixture Models (GMM), Self-Organizing Maps (SOM), Independent Component Analysis (ICA), Isomap (Isometric Mapping)
<b>UNIT-4</b>	<b>SEMI-SUPERVISED LEARNING TECHNIQUES</b>
<b>DAY-9</b>	Self-training Co-Training Semi-Supervised Support Vector Machines (S3VM)
<b>DAY-10</b>	Label Propagation Semi-Supervised Generative Adversarial Networks (SGANs) Pseudo-Labeling
<b>UNIT-5</b>	<b>REINFORCEMENT LEARNING TECHNIQUES</b>
<b>DAY-11</b>	Introduction to Reinforcement Learning (RL), Key components of RL: agent, environment, actions, rewards, and policies
<b>DAY-12</b>	Markov Decision Processes (MDPs): Definition and components of MDPs, Markov property and the notion of state transitions Value functions, Bellman equations and optimality conditions
<b>DAY-13</b>	Dynamic programming algorithms: Policy Evaluation, Policy Improvement, Policy Iteration, and Value Iteration
<b>DAY-14</b>	Monte Carlo Methods: Introduction, Monte Carlo Prediction and Control, Exploration and Exploitation trade-offs

## **UNIT-6**

### **ADVANCED MACHINE LEARNING TECHNIQUES**

- DAY-15** Ensemble Learning: GBMs and XGBoost
- DAY-16** Dimensionality Reduction Techniques: t-SNE, UMAP
- DAY-17** Model Evaluation and Hyperparameter Tuning
- DAY-18** Case Studies: Advanced ML Applications in Healthcare, Finance, and Marketing

#### **PRACTICAL SESSION-1**

- DAY-19** Implementing linear regression using scikit-learn
- DAY-20** Implementing K-Means clustering algorithm using scikit-learn
- DAY-21** Implementing self-training algorithm for semi-supervised classification, Pseudo-labeling unlabeled data and incorporating it into the training set
- DAY-21** Implementing a simple RL environment and agent interaction loop in Python

#### **PRACTICAL SESSION-2**

- DAY-22** Training ensemble models on a sample dataset and evaluating performance
- DAY-23** Implementing t-SNE and UMAP algorithms for dimensionality reduction
- DAY-24** Performing model evaluation using cross-validation techniques, Selecting the best model based on evaluation metrics and hyperparameter optimization results

## STRUCTURE OF DL COURSE

**SUBJECT NAME**

### **3. DEEP LEARNING**

<b>UNIT-1</b>	<b>FUNDAMENTALS OF DEEP LEARNING</b>
<b>DAY-1</b>	Introduction to Deep Learning, Neural Network Architectures: Perceptrons, Multi Layer Perceptrons, Feed Forward Neural Networks
<b>DAY-2</b>	Activation Functions, Unit saturation, vanishing gradient problem, Heuristics for faster training, Nestors accelerated gradient descent, Regularization, Dropout
<b>DAY-3</b>	Deep Learning Frameworks: TensorFlow, Keras, PyTorch, Keras, MXNet, Caffe, Chainer, TensorFlow.js
<b>UNIT-2</b>	<b>CONVOLUTIONAL NEURAL NETWORKS (CNNs)</b>
<b>DAY-4</b>	CNN Architecture and Components- convolution/pooling layers, Image Classification with CNNs
<b>DAY-5</b>	Transfer Learning and Fine-Tuning Pre-trained CNNs, Sentence Classification using Convolutional Neural Networks
<b>DAY-6</b>	Advanced CNN Architectures: ResNet, Inception, EfficientNet
<b>UNIT-3</b>	<b>RECURRENT NEURAL NETWORKS (RNNs)</b>
<b>DAY-7</b>	Basics of RNNs, Applications of RNNs in NLP and Time Series Analysis, Training and Optimization Techniques for RNNs
<b>DAY-8</b>	Recurrent Neural Networks: LSTM, GRU, Encoder Decoder architectures.
<b>DAY-9</b>	Analogy Reasoning: Named Entity Recognition, Opinion Mining using Recurrent Neural Networks, Dialogue Generation with LSTMs
<b>UNIT-4</b>	<b>FAST RCNN, FASTER RCNN AND YOLO MODELS</b>
<b>DAY-10</b>	Introduction to Object Detection, Object Detection Techniques, R- CNN Drawbacks of Regional Convolutional Neural Networks, FAST R-CNN Architecture, Moving from FAST R-CNN to FASTER R-CNN.
<b>DAY-11</b>	Introduction to YOLO Models, Drawbacks of R-CNN and its variants, YOLO Model Examples, YOLO Versions v1 through v7
<b>UNIT-5</b>	<b>DEEP UNSUPERVISED LEARNING</b>
<b>DAY-12</b>	Introduction to Adversarial Generative Networks, Applications of GANs: Image Generation, Style Transfer
<b>DAY-13</b>	Introduction to Variational Auto Encoders, Applications of VAEs: Image Reconstruction, Anomaly Detection,
<b>DAY-14</b>	Auto-encoder and DBM Attention and memory models, Dynamic Memory Models
<b>UNIT-6</b>	<b>DL APPLICATIONS AND CASE STUDIES</b>
<b>DAY-15</b>	Applications of Deep Learning to Computer Vision: Image segmentation, object detection, automatic image captioning, Image generation with Generative adversarial networks, video to text with

LSTM models, Attention Models for computer vision tasks

**DAY-16** Applications of Deep Learning to NLP: Introduction to NLP and Vector Space Model of Semantics, Word Vector Representations: Continuous Skip-Gram Model

**DAY-17** Continuous Bag-of-Words model (CBOW), Glove, Evaluations and Applications in word similarity

**DAY-18** Deep Learning in Healthcare: Medical Imaging, Disease Diagnosis  
Deep Learning in Finance: Algorithmic Trading, Fraud Detection

### **PRACTICAL SESSION-1**

**DAY-19** Implementing deep learning models for computer vision tasks (image segmentation, object detection, image captioning)

**DAY-20** Implementing transfer learning for image classification and sentence classification

Generating dialogue responses with LSTM models

**DAY-21** Implementing adversarial generative networks (GANs) for image generation

Implementing basic RNN architectures for sequential data analysis

### **PRACTICAL SESSION-2**

**DAY-22** Applying variational autoencoders (VAEs) for image reconstruction and anomaly detection

**DAY-23** Implementing R-CNN, Fast R-CNN and Faster R-CNN architectures for object detection

Implementation of YOLO Models for Object Detection

**DAY-24** Applying deep learning techniques to NLP tasks (word vector representations, word similarity evaluation)

# REFERENCES

## **Artificial Intelligence (AI):**

*"Artificial Intelligence: A Modern Approach"* by Stuart Russell and Peter Norvig

*"Artificial Intelligence: Foundations of Computational Agents"* by David L. Poole and Alan K. Mackworth

*"Artificial Intelligence: Structures and Strategies for Complex Problem Solving"* by George F. Luger

*"Pattern Recognition and Machine Learning"* by Christopher M. Bishop

*"Artificial Intelligence: A Guide to Intelligent Systems"* by Michael Negnevitsky

## **Machine Learning (ML):**

*"Machine Learning: A Probabilistic Perspective"* by Kevin P. Murphy

*"Introduction to Machine Learning"* by Ethem Alpaydin

*"Pattern Recognition and Machine Learning"* by Christopher M. Bishop

*"Deep Learning"* by Ian Goodfellow, Yoshua Bengio, and Aaron Courville

*"Understanding Machine Learning: From Theory to Algorithms"* by Shai Shalev-Shwartz and Shai Ben-David

## **Deep Learning (DL):**

*"Deep Learning"* by Ian Goodfellow, Yoshua Bengio, and Aaron Courville

*"Deep Learning for Computer Vision"* by Rajalingappaa Shanmugamani

*"Deep Learning: A Practitioner's Approach"* by Adam Gibson and Josh Patterson

*"Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow"* by Aurélien Géron

*"Neural Networks and Deep Learning: A Textbook"* by Charu C. Aggarwal

## **Additional References:**

*"Reinforcement Learning: An Introduction"* by Richard S. Sutton and Andrew G. Barto

*"Natural Language Processing with Python"* by Steven Bird, Ewan Klein, and Edward Loper

*"Bayesian Reasoning and Machine Learning"* by David Barber

*"The Hundred-Page Machine Learning Book"* by Andriy Burkov

*"Data Science for Business: What You Need to Know about Data Mining and Data-Analytic Thinking"* by Foster Provost and Tom Fawcett