# 2-Month Data Science & Machine Learning Training Program

# Applications-Driven Approach for Career Development

# **Program Overview**

• Duration: 8 weeks (2 months)

• **Schedule:** 3 sessions per week × 3 hours per session

• Total Hours: 72 hours

• Target Audience: Graduate students seeking industry-ready skills

• Approach: Project-based learning with portfolio development

• Outcome: Job-ready skills with 8-10 completed projects on GitHub — since you want to **teach with** more emphasis on applications and already have this curated **GitHub repo of projects**, we can build your 2-month, 24-session training outline **around these projects** instead of abstract theory. This way, the student gets **portfolio-ready work** while learning the concepts.

Here's a refined 8-week (24 sessions) plan using the repo projects:

# 2-Month Data Science & ML Training (Applications-Driven)

Format: 3x/week × 3 hrs x 8 weeks Focus: Hands-on projects → GitHub portfolio

# **Detailed Weekly Curriculum**

# Week 1: Foundations & Exploratory Data Analysis (EDA)

Focus: Building strong Python foundations for data science

Session 1 (3 hours): Python Fundamentals & Environment Setup

## **Topics:**

- Python refresher: data types, control structures, functions
- Jupyter notebook setup and best practices
- Introduction to data science libraries ecosystem
- Git and GitHub setup for portfolio development

- Set up development environment (Anaconda, Jupyter, VS Code)
- Create GitHub repository structure
- Basic Python exercises with data structures

#### Session 2 (3 hours): Data Manipulation with Pandas & NumPy

#### **Topics:**

- Pandas DataFrames: creation, indexing, filtering, grouping
- NumPy arrays: mathematical operations, broadcasting
- Data loading from various sources (CSV, JSON, APIs)
- Basic data cleaning techniques

#### **Practical Activities:**

- Load and explore multiple datasets
- Practice data manipulation exercises
- Introduction to the Iris Dataset

### Session 3 (3 hours): Data Visualisation & EDA Techniques

# **Topics:**

- Matplotlib fundamentals: plots, subplots, customisation
- Seaborn for statistical visualisation
- EDA methodology: univariate, bivariate, multivariate analysis
- Statistical summaries and distributions

#### **Project: Iris Dataset Classification**

- Complete EDA with visualisations
- Build first ML model (basic classification)
- Document findings and push to GitHub

Week 1 Deliverable: Comprehensive Jupyter notebook with EDA + ML model on GitHub

# Week 2: Data Preprocessing & Regression Analysis

Focus: Advanced data cleaning and regression modelling

#### Session 4 (3 hours): Advanced Data Cleaning & Feature Engineering

#### **Topics:**

- Handling missing values: imputation strategies
- · Outlier detection and treatment
- Data type conversions and categorical encoding
- Feature scaling and normalisation techniques

- Work with messy datasets
- · Implement various imputation methods
- Create feature engineering pipeline

#### Session 5 (3 hours): Regression Fundamentals & Implementation

#### **Topics:**

- Linear regression theory and assumptions
- Multiple regression and polynomial features
- Regularisation techniques (Ridge, Lasso, Elastic Net)
- Model evaluation metrics for regression

#### **Project Start: House Price Regression**

- Load and explore housing dataset
- Feature engineering for price prediction
- Implement multiple regression models

#### Session 6 (3 hours): Advanced Regression & Business Applications

# **Topics:**

- Cross-validation techniques
- Hyperparameter tuning with GridSearch
- Feature selection methods
- Business interpretation of model results

#### **Projects:**

- Complete House Price Regression with model tuning
- Start BigMart Sales Prediction project
- Compare different regression approaches

Week 2 Deliverable: Two complete regression projects with documentation and model comparison

# Week 3: Classification Algorithms & Evaluation

Focus: Binary and multi-class classification problems

#### Session 7 (3 hours): Classification Fundamentals

#### **Topics:**

- Logistic regression: theory and implementation
- · Decision trees: splitting criteria, pruning
- Classification vs regression: key differences
- Probability interpretation in classification

#### **Project Start: Loan Prediction**

- Explore loan approval dataset
- Implement logistic regression
- Feature importance analysis

#### Session 8 (3 hours): Ensemble Methods & Advanced Classification

#### **Topics:**

- Random Forest: bagging and feature randomness
- Gradient Boosting and XGBoost
- Ensemble voting methods
- Handling imbalanced datasets

#### **Practical Activities:**

- Compare individual vs ensemble models
- Implement SMOTE for imbalanced data
- Cross-validation for model selection

#### Session 9 (3 hours): Model Evaluation & Business Metrics

# **Topics:**

- Confusion matrix and classification report
- ROC curves and AUC interpretation
- Precision, recall, F1-score trade-offs
- Business-specific evaluation metrics

#### **Project: Titanic Survival Prediction**

- Complete end-to-end classification pipeline
- Comprehensive model evaluation
- Feature engineering for survival prediction

Week 3 Deliverable: Two classification projects with comprehensive evaluation metrics and business insights

# Week 4: Natural Language Processing (NLP)

Focus: Text data processing and sentiment analysis

#### Session 10 (3 hours): Text Preprocessing & NLP Fundamentals

#### **Topics:**

- Text cleaning: tokenisation, stemming, lemmatisation
- Stop words removal and text normalisation
- Regular expressions for text processing
- Introduction to NLTK and spaCy libraries

- Process raw text data from various sources
- Build text preprocessing pipeline
- Explore Twitter and news datasets

#### Session 11 (3 hours): Feature Extraction from Text

#### **Topics:**

- Bag of Words (BoW) model
- TF-IDF (Term Frequency-Inverse Document Frequency)
- N-grams and character-level features
- Word embeddings introduction

#### **Project Start: Sentiment Analysis**

- Amazon reviews dataset exploration
- Implement BoW and TF-IDF approaches
- Compare feature extraction methods

#### Session 12 (3 hours): NLP Classification & Applications

#### **Topics:**

- Naive Bayes for text classification
- Support Vector Machines for text
- Model evaluation for NLP tasks
- Handling multi-class text classification

#### **Projects:**

- Complete Twitter Sentiment Analysis
- Fake News Detection implementation
- Email Classification (spam vs ham)

Week 4 Deliverable: Three NLP projects demonstrating different text classification approaches

# Week 5: Introduction to Deep Learning

Focus: Neural networks and computer vision basics

#### Session 13 (3 hours): Neural Network Fundamentals

#### **Topics:**

- Perceptron and multi-layer perceptrons
- Backpropagation algorithm
- Activation functions and their properties
- Introduction to TensorFlow/Keras

- Build neural network from scratch
- Implement simple MLP with Keras
- Understand gradient descent visualisation

#### Session 14 (3 hours): Convolutional Neural Networks (CNNs)

#### **Topics:**

- CNN architecture: convolution, pooling, fully connected layers
- Feature maps and filters
- CNN for image classification
- Transfer learning concepts

#### **Project: Handwritten Digit Recognition (MNIST)**

- Load and preprocess MNIST dataset
- Build CNN from scratch
- Compare with traditional ML approaches

#### Session 15 (3 hours): Deep Learning Best Practices

# **Topics:**

- Overfitting prevention: dropout, early stopping
- Batch normalisation and optimisation
- Model architecture design principles
- Hyperparameter tuning for deep learning

#### **Project: Wine Quality Prediction**

- Apply deep learning to tabular data
- Compare neural networks with ensemble methods
- Document model architecture decisions

Week 5 Deliverable: Two deep learning projects with architecture documentation and performance analysis

# **Week 6: Computer Vision Applications**

Focus: Advanced image processing and real-world CV applications

#### Session 16 (3 hours): OpenCV Fundamentals & Image Processing

#### **Topics:**

- Image loading, manipulation, and basic operations
- Colour spaces and transformations
- Edge detection and contour analysis
- Basic image filtering and morphological operations

#### **Project: Color Detection**

- Implement colour detection system
- · Real-time colour tracking with webcam
- HSV colour space applications

#### Session 17 (3 hours): Advanced CNN Architectures & Transfer Learning

#### **Topics:**

- Pre-trained models: VGG, ResNet, MobileNet
- Transfer learning strategies
- Fine-tuning vs feature extraction
- Model optimisation for deployment

#### **Project: Traffic Signs Recognition**

- Use German Traffic Sign Recognition Benchmark
- Implement transfer learning with pre-trained CNN
- Real-time traffic sign detection

#### Session 18 (3 hours): Real-time Computer Vision Applications

# **Topics:**

- Facial landmark detection
- Real-time video processing
- Integration of deep learning with OpenCV
- Model deployment considerations

# **Project: Driver Drowsiness Detection**

- Eye blink detection using facial landmarks
- Real-time drowsiness alert system
- Combine OpenCV with deep learning models

**Week 6 Deliverable:** Three computer vision applications with real-time capabilities and deployment documentation

# Week 7: Machine Learning Systems & Recommendation Engines

Focus: Unsupervised learning and production ML systems

#### Session 19 (3 hours): Unsupervised Learning & Clustering

#### **Topics:**

- K-means clustering algorithm and implementation
- · Hierarchical clustering methods
- DBSCAN for density-based clustering
- Dimensionality reduction with PCA

# **Project: Customer Segmentation**

- RFM analysis for customer segmentation
- K-means clustering implementation

• Business insights from clustering results

#### Session 20 (3 hours): Recommendation Systems

#### **Topics:**

- Collaborative filtering vs content-based filtering
- Matrix factorisation techniques
- Similarity metrics and nearest neighbours
- Evaluation metrics for recommendation systems

### **Project Start: Movie Recommendation System**

- Implement collaborative filtering
- Content-based filtering with movie features
- Hybrid recommendation approach

#### Session 21 (3 hours): Anomaly Detection & Production Systems

#### **Topics:**

- Statistical methods for anomaly detection
- Isolation Forest and One-Class SVM
- Model deployment with Streamlit/Gradio
- API development for ML models

# **Projects:**

- Complete Movie Recommendation System with web interface
- Credit Card Fraud Detection using anomaly detection
- Deploy recommendation system as web application

Week 7 Deliverable: Deployed web applications for recommendation system and fraud detection

# **Week 8: Capstone Project & Career Preparation**

**Focus:** End-to-end ML project and job readiness

#### Session 22 (3 hours): Capstone Project Planning & Advanced Applications

#### **Topics:**

- · Project planning and requirement analysis
- Advanced deep learning applications
- Medical image analysis with CNNs
- Transfer learning for specialised domains

## **Capstone Project Options:**

- Covid-19 X-ray Classification (medical imaging)
- Human Activity Recognition (time series + sensors)

• Custom project based on student interest

#### Session 23 (3 hours): Model Development & Evaluation

#### **Topics:**

- End-to-end ML pipeline development
- Model versioning and experiment tracking
- Advanced evaluation techniques
- A/B testing concepts for ML

#### **Activities:**

- Develop complete capstone project
- Implement comprehensive evaluation framework
- Document methodology and results

# Session 24 (3 hours): Portfolio Optimization & Career Preparation

### **Topics:**

- GitHub portfolio structuring and documentation
- Technical resume optimisation
- LinkedIn profile enhancement
- Interview preparation and mock interviews

#### **Career Activities:**

- Portfolio review and optimisation
- README documentation for all projects
- Resume and LinkedIn profile updates
- Mock technical interviews with feedback

Week 8 Deliverable: Complete portfolio with capstone project + career preparation materials

# Project-to-Week Mapping Summary

| Week | Repository Projects Used  | Difficulty Level      |
|------|---|-----------------------|
| 1    | Iris Dataset Classification   | Basic                 |
| 2    | House Price Regression, BigMart Sales Dataset                           | Basic                 |
| 3    | Loan Prediction, Titanic Dataset  | Basic                 |
| 4    | Sentiment Analysis, Fake News Detection, Email Classification           | Basic/Intermediate    |
| 5    | Handwritten Digit Recognition, Wine Quality                             | Intermediate          |
| 6    | Color Detection, Traffic Signs Recognition, Driver Drowsiness Detection | Intermediate/Advanced |

| Week | Repository Projects Used  | Difficulty Level |
|------|---|------------------|
| 7    | Customer Segmentation, Song/Movie Recommendation, Credit Card Fraud Detection | Advanced         |
| 8    | Covid-19 X-ray Classification, Human Activity Recognition                     | Advanced         |

# Session-by-Session Breakdown

| Session | Week | Duration  | Learning Objectives                            | Practical Activities                            | Assessment                             |
|---------|------|-----------|--|---|--|
| 1-3     | 1    | 3hrs each | Python DS<br>fundamentals, EDA<br>techniques   | Iris dataset exploration and classification     | Code review +<br>project<br>submission |
| 4-6     | 2    | 3hrs each | Data preprocessing and regression modeling     | House price prediction with feature engineering | Model<br>performance<br>evaluation     |
| 7-9     | 3    | 3hrs each | Classification algorithms and metrics          | Binary and multi-class classification projects  | Confusion matrix and ROC analysis      |
| 10-12   | 4    | 3hrs each | Natural Language<br>Processing<br>fundamentals | Text classification and sentiment analysis      | NLP pipeline implementation            |
| 13-15   | 5    | 3hrs each | Deep learning and neural networks              | CNN implementation for image classification     | Model<br>architecture<br>documentation |
| 16-18   | 6    | 3hrs each | Computer vision applications                   | Real-time image processing with OpenCV          | Live demo + code<br>deployment         |
| 19-21   | 7    | 3hrs each | Unsupervised learning and systems              | Recommendation system with web interface        | Functional web application             |
| 22-24   | 8    | 3hrs each | Capstone project and career preparation        | End-to-end ML project + portfolio review        | Final presentation + portfolio         |

# Learning Outcomes & Portfolio Components

| Component            | Description   | Industry Relevance                      | Portfolio Impact                           |
|----------------------|---|---|--|
| Technical Skills     | Python, pandas, scikit-learn,<br>TensorFlow/PyTorch, OpenCV | Direct application in DS/ML roles       | Demonstrates coding proficiency            |
| Project<br>Portfolio | 8-10 complete projects with documentation                   | Shows practical problem-solving ability | Differentiates from theoretical candidates |
| Domain<br>Knowledge  | Finance, healthcare, NLP, computer vision applications      | Versatility across industries           | Appeals to diverse employers               |

| Component  | Description                                 | Industry Relevance                      | Portfolio Impact                    |  |
|--|---|---|-------------------------------------|--|
| Deployment<br>Skills   | Streamlit, Gradio, GitHub Pages integration | End-to-end<br>development<br>capability | Shows production-                   |  |
| Communication Technical documentation, README files, presentations |   | Essential for team collaboration        | Demonstrates professional readiness |  |

# Assessment & Evaluation Framework

| <b>Evaluation Type</b>         | Weight | Frequency | Criteria                                      |
|--------------------------------|--------|-----------|---|
| <b>Project Submissions</b>     | 60%    | Weekly    | Code quality, documentation, results analysis |
| <b>Technical Presentations</b> | 20%    | Bi-weekly | Communication skills, technical depth         |
| Peer Code Reviews              | 10%    | Ongoing   | Collaboration, feedback incorporation         |
| Final Capstone                 | 10%    | Week 8    | End-to-end project execution                  |

# **Career Preparation Components**

| Activity                  | Timeline     | Objective                       | Deliverable                                   |
|---------------------------|--------------|---------------------------------|---|
| GitHub Portfolio<br>Setup | Week 1-<br>2 | Professional online presence    | Structured repository with README             |
| LinkedIn<br>Optimization  | Week 6       | Network building and visibility | Updated profile with project highlights       |
| Resume<br>Enhancement     | Week 7       | Job application readiness       | Technical resume with quantified achievements |
| Interview<br>Preparation  | Week 8       | Technical interview skills      | Mock interview performance + feedback         |
| Industry<br>Networking    | Ongoing      | Professional connections        | LinkedIn engagement + community participation |

# **Expected Outcomes**

Upon completion of this program, students will have:

- Technical Portfolio: 8-10 documented projects demonstrating ML/DS capabilities
- Industry Skills: Proficiency in Python ecosystem and modern ML frameworks
- Practical Experience: End-to-end project development from data to deployment
- Career Materials: Professional GitHub profile, optimized resume, and interview skills
- Domain Expertise: Applied knowledge across multiple industries and problem types