# Foundations of Machine Learning

By

Dr Ravi Prakash Verma Professor

Department of CSAI

**ABESIT** 

## Glimpse

- Total Duration: 39+ Hours (13 Days)
- Mode: Hands-on coding in Python (NumPy, Pandas, Scikit-learn, Matplotlib, Seaborn).
- Target Audience: Students
- Prerequisites: Python programming, pandas, mathplot, numpy and fundamental statistics knowledge.

## **Objectives**

- By the end of this workshop, participants will:
  - Understand theoretical concepts behind key ML models.
  - Implement models using Python and Scikit-learn.
  - Analyze real-world datasets for predictive modeling.
  - Optimize models using feature selection and hyperparameter tuning.
  - **Evaluate models using appropriate performance metrics (R<sup>2</sup>, RMSE, MAE, ROC-AUC, etc.).**

## **Workshop Structure & Duration**

Day	Topics Covered
Day 1	Introduction to ML & Data Preprocessing
Day 2	Linear Regression - Concept & Mathematics
Day 3	Linear Regression in Python
Day 4	Model Performance Metrics (R², RMSE, MAE, MSE)
Day 5	Confusion Matrix & Model Performance
Day 6	Logistic Regression - Concept & Mathematics
Day 7	Project on Logistic Regression in Python

## **Workshop Structure & Duration**

Day 8	Decision Trees - Concept & Theory
Day 9	Project on Decision Trees in Python
Day 10	Support Vector Machines (SVM) - Linear SVM
Day 11	Support Vector Machines (SVM) - Non-Linear SVM
Day 12	Project on Linear & Non-Linear SVM in Python
Day 13	Future Learning Path (Deep Learning, NLP, etc.)

- Day 1: Introduction & Data Preprocessing (3 Hours)
- Theory:
  - Overview of Supervised vs. Unsupervised Learning
  - Importance of Feature Engineering
  - Handling missing values, outliers, and categorical variables
  - Data normalization & standardization
- Hands-on Coding:
  - Importing datasets using Pandas
  - Handling missing data using mean/median imputation
- Feature scaling using MinMaxScaler & StandardScaler

Day 2: Linear Regression - Concept & Mathematics (3 Hours)

#### Theory:

- Introduction to Regression Problems
- Mathematical formulation of Simple & Multiple Linear Regression
- Understanding Cost Function & Gradient Descent

- Implementing Linear Regression from scratch using NumPy
- Visualizing best-fit line using Matplotlib

- Day 3: Linear Regression in Python (3 Hours)
- Theory:
  - Feature Selection & Model Complexity
  - Handling Multicollinearity using VIF
- Hands-on Coding:
  - Using Scikit-learn's LinearRegression
- Implementing Feature Selection Techniques

- Day 4: Model Performance Metrics (R<sup>2</sup>, RMSE, MAE, MSE) (3 Hours)
- Theory:
- Understanding R<sup>2</sup> (Coefficient of Determination)
- Root Mean Squared Error (RMSE) vs. Mean Absolute Error (MAE)
- Importance of Adjusted R<sup>2</sup> for multiple regression
- Hands-on Coding:
- Implementing R<sup>2</sup>, RMSE, MAE, and MSE in Python
- Comparing models using different metrics
- Confusion Matrix & Model Performance
  - o True Positive (TP), True Negative (TN), False Positive (FP), False Negative (FN)
  - o Precision, Recall, and F1-Score
  - Visual Representation of Confusion Matrix
  - Code Implementation for Confusion Matrix
- ROC & AUC Score
  - Understanding ROC Curve
  - Computing AUC Score for Model Performance

- Day 5: Confusion Matrix & Model Performance
- . Confusion Matrix & Model Performance
  - True Positive (TP), True Negative (TN), False Positive (FP), False Negative (FN)
  - Precision, Recall, and F1-Score
  - Visual Representation of Confusion Matrix
  - Code Implementation for Confusion Matrix
- . ROC & AUC Score
  - . Understanding ROC Curve
  - Computing AUC Score for Model Performance

- Day 6: Logistic Regression Concept & Mathematics (3 Hours)
- Theory:
  - Difference between Regression & Classification
  - Sigmoid Function & Decision Boundary
- . Understanding Cost Function for Logistic Regression
- Hands-on Coding:
  - Implementing Sigmoid Function from scratch
  - Plotting decision boundaries for classification

- Day 7: Project on Logistic Regression in Python (3 Hours)
- Theory:
  - Confusion Matrix, Precision, Recall, F1-Score
  - Introduction to ROC & AUC for classification
- Hands-on Coding:
  - Implementing Logistic Regression using Scikit-learn
- Plotting ROC Curve & AUC Score

Day 8: Decision Trees - Concept & Theory (3 Hours)

## Theory:

- Understanding Entropy, Information Gain, and Gini Impurity
- Overfitting in Decision Trees
- Pruning Techniques

- Implementing Decision Tree from scratch
- Visualizing Tree Structure using Graphviz

Day 9: Project on Decision Trees in Python (3 Hours)

#### Theory:

- Feature Importance in Decision Trees
- Hyperparameter Tuning using GridSearchCV

- . Implementing DecisionTreeClassifier using Scikit-learn
- Optimizing decision trees for better accuracy

Day 10: Linear SVM - Concept & Theory (3 Hours)

## Theory:

- Introduction to Support Vector Machines (SVM)
- Understanding Hyperplanes & Support Vectors
- Hard Margin vs. Soft Margin SVM

- Implementing Linear SVM using Scikit-learn
- Visualizing decision boundaries in 2D classification problems

Day 11: Non-Linear SVM - Concept & Theory (3 Hours)

## • Theory:

- Understanding Kernel Trick for Non-Linear Classification
- . Types of Kernels: Polynomial, Gaussian (RBF), Sigmoid
- Tuning SVM Parameters (C, Gamma, Kernel Choice)

- Implementing Non-Linear SVM using RBF Kernel
- . Visualizing Complex Decision Boundaries

 Day 12: Project on Linear & Non-Linear SVM in Python (3 Hours)

## Theory:

- Comparing Linear vs. Non-Linear SVM Performance
- Choosing the Right Kernel for Different Datasets

- . Training Linear and Non-Linear SVM models
- . Hyperparameter tuning using GridSearchCV

- Day 13: Future Learning Path
- Theoretical discussion on
  - (Deep Learning, NLP, etc.)

## **Learning Outcomes**

- By the end of this workshop, participants will:
  - Master Linear Regression, Logistic Regression, Decision Trees, and SVM.
  - Implement both Linear & Non-Linear SVM with different kernels.
  - Optimize models using feature selection & hyperparameter tuning.
  - Evaluate models using R², RMSE, MAE, ROC-AUC, and more.