

# Certificate Program in Data Science

The following is the detailed syllabus for the certificate program offered at our institute:

## 1. PYTHON FOR DATA SCIENCE

### *Module 1: Python Intro and Basics*

- ▣ Python Intro and Get Started
- ▣ Python Syntax and Comments
- ▣ Variables and Data types
- ▣ Operators and types

### *Module 2: Python Data Structures*

- ▣ String and String methods
- ▣ List: Object, methods, comprehensions
- ▣ Tuple: Object, methods, comprehensions
- ▣ Sets: Object, methods, comprehensions
- ▣ Dictionary: Object, methods, comprehensions

### *Module 3: Conditional Operators*

- ▣ If Else statement
- ▣ Nested If
- ▣ For Loops
- ▣ While Loops
- ▣ Break and Continue statements

### *Module 4: Python Functions and Classes*

- ▣ Python Functions Intro
- ▣ Creating and Calling Functions
- ▣ Types of Function arguments
- ▣ Lambda Functions
- ▣ Pass statement

### *Module 5: Numpy Library*

- ▣ Numpy Intro
- ▣ Arrays – 1D, 2D and 3D
- ▣ Array Creation
- ▣ Array Manipulation
- ▣ Matrix Operations
- ▣ Numpy Random Module
- ▣ Arrays – Stacking, Splitting & Slicing

- ▣ Broadcasting and Boolean Mask
- ▣ Other Functions

### ***Module 6: Pandas Library***

- ▣ Pandas Intro
- ▣ Pandas Series and DataFrame
- ▣ Reading the Data
- ▣ DataFrame methods

### ***Module 7: Python Visualization Library***

- ▣ About Matplotlib, Seaborn libraries
- ▣ Plot, subplot creation
- ▣ Basic Plots: line, bar, scatter, pie charts
- ▣ multi-plot, pair plots
- ▣ Plotting with Seaborn

## **2. STATISTICS FOR DATA SCIENCE**

- ⦿ Descriptive And Inferential statistics
- ⦿ Types Of Data, Sampling types
- ⦿ Measures of Central Tendencies
- ⦿ Data Variability: Standard Deviation
- ⦿ Z-Score, Outliers, Normal Distribution
- ⦿ Central Limit Theorem
- ⦿ Histogram, Normality Tests
- ⦿ Skewness & Kurtosis
- ⦿ Understanding Hypothesis Testing
- ⦿ P-Value Method, Types Of Errors
- ⦿ T Distribution, One Sample T-Test
- ⦿ Independent And Relational T Tests
- ⦿ Direct And Indirect Correlation
- ⦿ Regression Theory

## **3. MACHINE LEARNING**

### ***Module 1: Machine Learning Introduction***

- ▶ Machine Learning Introduction
- ▶ ML core concepts
- ▶ Supervised and Unsupervised Learning
- ▶ Reinforcement Learning Intro
- ▶ Regression and Classification Algorithms`

## ***Module 2: Linear Regression (Hands-on)***

- ▶ About Linear Regression
- ▶ Mathematical Intuition
- ▶ Cost Function Minimization Overview
- ▶ Model Evaluation Metrics: MSE, MAE, RMSE etc.

## ***Module 3: Logistic Regression (Hands-on)***

- ▶ Logistic Regression – Introduction
- ▶ Sigmoid Function
- ▶ Cost Function Minimization Overview
- ▶ Model Evaluation Metrics: Accuracy, Precision, Recall etc.

## ***Module 4: K Nearest Neighbour Classifier (KNN)***

- ▶ KNN Intro and working
- ▶ Nearest Neighbour concepts
- ▶ Model Evaluation Metrics

## ***Module 5: K Means Clustering***

- ▶ Clustering Introduction
- ▶ Algorithm Working
- ▶ Cost Function Minimization Overview
- ▶ Model Evaluation Metrics

## ***Module 6: Decision Trees***

- ▶ Regression and Classification Trees
- ▶ Ensemble Techniques: Bagging and Boost
- ▶ Random Forest and XG Boost Algorithms
- ▶ Cost Function Minimization Overview
- ▶ Model Evaluation Metrics

## ***Module 7: Naive Bayes Classifier***

- ▶ Introduction to Naive Bayes
- ▶ How it works: Bayes' Theorem
- ▶ Naive Bayes For Text Classification
- ▶ Modelling and Evaluation in Python

## ***Module 8: Support Vector Machine (SVM)***

- ▶ About SVM Classifier
- ▶ Support Vectors and Hyperplanes
- ▶ Linear and Non-linear SVMs
- ▶ Kernel methods in SVM

### ***Module 9: Artificial Neural Network***

- ▶ Introduction to ANN
- ▶ Modelling and Evaluation of ANN in Python
- ▶ Image Classification using CNN

### ***Module 10: Time Series Forecasting***

- ▶ Time Series Intro and Data
- ▶ Trend, Seasonality, cyclical and random
- ▶ Autoregressive Model (AR)
- ▶ Moving Average Model (MA)
- ▶ Stationarity of Time Series
- ▶ ARIMA Model
- ▶ Autocorrelation and AIC

### ***Module 11: Principal Component Analysis (PCA)***

- ▶ About PCA
- ▶ Working and Methodology of PCA
- ▶ Modelling PCA in Python

### ***Module 12: Other ML Concepts***

- ▶ Adv. Metrics (Roc\_Auc, R2, Precision, Recall)
- ▶ K-Fold Cross-validation
- ▶ Hyper Parameter Tuning
- ▶ Grid And Randomized Search CV
- ▶ Balancing Imbalanced Data Set
- ▶ Feature Selection Techniques

### ***Module 13: Feature Engineering***

- ▶ Introduction to Feature Engineering
- ▶ Transforming Predictors
- ▶ Feature Selection methods
- ▶ Backward elimination technique
- ▶ Feature importance from ML modelling

### ***Module 14: ML Model Deployment***

- ▶ Introduction to Flask, Streamlit
- ▶ URL and App routing
- ▶ Flask, Streamlit application – ML Model Deployment

## **4. GIT FOR DATA SCIENTISTS**

### ***Module 1: GitHub Repo***

- 📦 About GitHub and Repo
- 📦 Creating GitHub Account
- 📦 Create a new Git Repo
- 📦 Copy existing repo
- 📦 Various Functionalities

### ***Module 2: Git commands***

- 📦 Git pull
- 📦 Code commits
- 📦 Push to remote repo
- 📦 Checkout branch
- 📦 Merge branches, Pull requests
- 📦 Other commands

### ***Module 3: GIT***

- 📦 Purpose of Version Control
- 📦 GIT Introduction
- 📦 Terminologies
- 📦 Git workflow

## **5. PROJECT WORK**

The project work will be on real data set. The learners can choose their project theme from the available options.