



# Data Science With Python



**The Best Training Institute in Hyderabad**



## Datascience with Python

### Lesson 1

#### **Course Introduction :**

- 1.1 Course Introduction
- 1.2 Demo Jupyter Lab Walk - Through

### Lesson 2

#### **Introduction to Data Science :**

- 2.1 Learning Objectives
- 2.2 Data Science Methodology
- 2.3 From Business Understanding to Analytic Approach
- 2.4 From Requirements to Collection
- 2.5 From Understanding to Preparation
- 2.6 From Modeling to Evaluation
- 2.7 From Deployment to Feedback
- 2.8 Key Takeaways



## Lesson 3

### **Python Libraries For Data Science :**

3.1 Learning Objectives

3.2 Python Libraries for Data Science

3.3 Import Library into Python Program

3.4 Numpy

3.5 Demo Numpy

3.6 Fundamentals of Numpy

3.7 Numpy Array Shapes and axes Part A

3.8 Numpy Array Shapes and axes Part

3.9 Arithmetic Operations

3.10 Conditional Statements in Python

3.11 Common Mathematical and Statistical Functions in NumPy

3.12 Indexing and Slicing in Python Part A

3.13 Indexing and Slicing in Python Part B

3.14 Introduction to Pandas

3.15 Introduction to Pandas Series

3.16 Querying a Series

3.17 Pandas Dataframe



- 3.18 Introduction to Pandas Panel
- 3.19 Common Functions in Pandas
- 3.20 Statistical Functions in Pandas
- 3.21 Date and Timedelta
- 3.22 IO Tools
- 3.23 Categorical Data
- 3.24 Working with Text Data
- 3.25 Iteration
- 3.26 Plotting with Pandas
- 3.27 Matplotlib
- 3.28 Demo Matplotlib
- 3.29 Data Visualization Libraries in Python Matplotlib
- 3.30 Graph Types
- 3.31 Using Matplotlib to Plot Graphs
- 3.32 Matplotlib for 3d Visualization
- 3.33 Using Matplotlib with Other Python Packages
- 3.34 Data Visualization Libraries in Python Seaborn An Introduction
- 3.35 Seaborn Visualization Features
- 3.36 Using Seaborn to Plot Graphs





3.37 Analysis using seaborn plots

3.38 Plotting 3D Graphs for Multiple Columns using Seaborn

3.39 SciPy

3.40 Demo Scipy

3.41 Scikit-learn

3.42 Scikit Models

3.43 Scikit Datasets

3.44 Preprocessing Data in Scikit Learn Part 1

3.45 Preprocessing Data in Scikit Learn Part 2

3.46 Preprocessing Data in Scikit Learn Part 3

3.47 Demo Scikit learn

3.48 Key Takeaways



## Lesson 4

### **Statistics :**

4.1 Learning Objectives

4.2 Introduction to Linear Algebra

4.3 Scalars and vectors

4.4 Dot product of Two Vectors

4.5 Linear Independence of Vectors

4.6 Norm of a Vector

4.7 Matrix

4.8 Matrix Operations

4.9 Transpose of a Matrix4.10 Rank of a Matrix

4.11 Determinant of a matrix and Identity matrix or operator

4.12 Inverse of a matrix and Eigenvalues and Eigenvectors

4.13 Calculus in Linear Algebra

4.14 Importance of Statistics for Data Scienc

4.15 Common Statistical Terms

4.16 Types of Statistics

4.17 Data Categorization and types of data

4.18 Levels of Measurement



- 4.19 Measures of central tendency mean
- 4.20 Measures of Central Tendency Median
- 4.21 Measures of Central Tendency Mode
- 4.22 Measures of Dispersion
- 4.23 Variance
- 4.24 Random Variables
- 4.25 Sets
- 4.26 Measure of Shape Skewness
- 4.27 Measure of Shape Kurtosis
- 4.28 Covariance and correlation
- 4.29 Basic Statistics with Python Problem Statement
- 4.30 Basic Statistics with Python Solution
- 4.31 Probability its Importance and Probability Distribution
- 4.32 Probability Distribution Binomial Distribution
- 4.33 Binomial Distribution using Python
- 4.34 Probability Distribution Poisson Distribution
- 4.35 Poisson Distribution Using Python
- 4.36 Probability Distribution Normal Distribution



- 4.37 Probability Distribution Uniform Distribution
- 4.38 Probability Distribution Bernoulli Distribution
- 4.39 Probability Density Function and Mass Function
- 4.40 Cumulative Distribution Function
- 4.41 Central Limit Theorem
- 4.42 Bayes Theorem
- 4.43 Estimation Theory
- 4.44 Point Estimate using Python
- 4.45 Distribution
- 4.46 Kurtosis Skewness and Student's T- distribution
- 4.47 Hypothesis Testing and mechanism
- 4.48 Hypothesis Testing Outcomes Type I and II Error
- 4.49 Null Hypothesis and Alternate Hypothesis
- 4.50 Confidence Intervals
- 4.51 Margin of Errors
- 4.52 Confidence Levels
- 4.53 T test and P values Using Python
- 4.54 Z test and P values Using Python
- 4.55 Comparing and Contrastin T test and Z-tests
- 4.56 Chi Squared Distribution



- 4.57 Chi Squared Distribution using Python
- 4.58 Chi squared Test and Goodness of Fit
- 4.59 ANOVA
- 4.60 ANOVA Terminologies
- 4.61 Assumptions and Types of ANOVA
- 4.62 Partition of Variance
- 4.63 F-distribution
- 4.64 F Distribution using Python
- 4.65 F-Test
- 4.66 Advanced Statistics with Python Problem Statement
- 4.67 Advanced Statistics with Python Solution
- 4.68 Key Takeaways





## Lesson 5

### **Data Wrangling :**

5.1 Learning Objectives

5.2 Data Exploration Loading Files Part A

5.3 Data Exploration Loading Files Part B

5.4 Data Exploration Techniques Part A

5.5 Data Exploration Techniques Part B

5.6 Seaborn

5.7 Demo Correlation Analysis

5.8 Data Wrangling

5.9 Missing Values in a Dataset

5.10 Outlier Values in a Dataset

5.11 Demo Outlier and Missing Value Treatment

5.12 Data Manipulation

5.13 Functionalities of Data Object in Python Part A

5.14 Functionalities of Data Object in Python Part B

5.15 Different Types of Joins

5.16 Key Takeaway



## Lesson 6

### **Feature Engineering :**

- 6.1 Learning Objectives
- 6.2 Introduction to Feature Engineering
- 6.3 Encoding of Catogorical Variables
- 6.4 Label Encoding
- 6.5 Techniques used for Encoding variables
- 6.6 Key Takeaways

## Lesson 7

### **Exploratory Data Analysis :**

- 7.1 Learning Objectives
- 7.2 Types of Plots
- 7.3 Plots and Subplots
- 7.4 Assignment 01 Pairplot Demo
- 7.5 Assignment 02 Pie Chart Demo
- 7.6 Key Takeaways



## Lesson 8

### **Feature Selection :**

8.1 Learning Objectives

8.2 Feature Selection

8.3 Regression

8.4 Factor Analysis

8.5 Factor Analysis Process

8.6 Key Takeaways



# Machine Learning

## Lesson 1

### Course Introduction :

- 1.1 Course Introduction
- 1.2 Demo Jupyter Lab Walk - Through

## Lesson 2

### Introduction to Machine Learning :

- 2.1 Learning Objectives
- 2.2 Relationship between Artificial Intelligence, Machine Learning, and Data Science: Part A
- 2.3 Relationship between Artificial Intelligence, Machine Learning, and Data Science: Part B
- 2.4 Definition and Features of Machine Learning
- 2.5 Machine Learning Approaches
- 2.6 Key Takeaways



## Lesson 3

### **Supervised Learning Regression and Classification :**

- 3.1 Learning Objectives
- 3.2 Supervised Learning
- 3.3 Supervised Learning: Real Life Scenario
- 3.4 Understanding the Algorithm
- 3.5 Supervised Learning Flow
- 3.6 Types of Supervised Learning: Part A
- 3.7 Types of Supervised Learning: Part B
- 3.8 Types of Classification Algorithms
- 3.9 Types of Regression Algorithms: Part A
- 3.10 Regression Use Case
- 3.11 Accuracy Metrics
- 3.12 Cost Function
- 3.13 Evaluating Coefficients
- 3.14 Demo: Linear Regression
- 3.15 Challenges in Prediction
- 3.16 Types of Regression Algorithms: Part B
- 3.17 Demo: Bigmart





- 3.18 Logistic Regression: Part A
- 3.19 Logistic Regression: Part B
- 3.20 Sigmoid Probability
- 3.21 Accuracy Matrix
- 3.22 Demo: Survival of Titanic Passengers
- 3.23 Overview of Classification
- 3.24 Classification: A Supervised Learning Algorithm
- 3.25 Use Cases
- 3.26 Classification Algorithms
- 3.27 Performance Measures: Confusion Matrix
- 3.28 Performance Measures: Cost Matrix
- 3.29 Naive Bayes Classifier
- 3.30 Steps to Calculate Posterior Probability: Part A
- 3.31 Steps to Calculate Posterior Probability: Part B
- 3.32 Support Vector Machines: Linear Separability
- 3.33 Support Vector Machines: Classification Margin
- 3.34 Linear SVM: Mathematical Representation
- 3.35 Non linear SVMs
- 3.36 The Kernel Trick
- 3.37 Demo: Voice Classification
- 3.38 Key Takeaways



## Lesson 4

### **Decision Trees and Random Forest :**

- 4.1 Learning Objectives
- 4.2 Decision Tree: Classifier
- 4.3 Decision Tree: Examples
- 4.4 Decision Tree: Formation
- 4.5 Choosing the Classifier
- 4.6 Overfitting of Decision Trees
- 4.7 Random Forest Classifier Bagging and Bootstrapping
- 4.8 Decision Tree and Random Forest Classifier
- 4.9 Demo: Horse Survival
- 4.10 Key Takeaways



## Lesson 5

### **Unsupervised Learning :**

5.1 Learning Objectives

5.2 Overview

5.3 Example and Applications of Unsupervised Learning

5.4 Clustering5.5 Hierarchical Clustering

5.6 Hierarchical Clustering: Example

5.7 Demo: Clustering Animals

5.8 K-means Clustering

5.9 Optimal Number of Clusters

5.10 Demo: Cluster Based Incentivization

5.11 Key Takeaways



## Lesson 6

### **Time Series Modelling :**

- 6.1 Learning Objectives
- 6.2 Overview of Time Series Modeling
- 6.3 Time Series Pattern Types: Part A
- 6.4 Time Series Pattern Types: Part B
- 6.5 White Noise
- 6.6 Stationarity
- 6.7 Removal of Non Stationarity
- 6.8 Demo: Air Passengers I
- 6.9 Time Series Models: Part A
- 6.10 Time Series Models: Part B
- 6.11 Time Series Models: Part C
- 6.12 Steps in Time Series Forecasting
- 6.13 Demo: Air Passengers II
- 6.14 Key Takeaways



## Lesson 7

### **Ensemble Learning :**

7.1 Learning Objectives

7.2 Overview

7.3 Ensemble Learning Methods: Part A

7.4 Ensemble Learning Methods: Part B

7.5 Working of AdaBoost

7.6 AdaBoost Algorithm and Flowchart

7.7 Gradient Boosting

7.8 XGBoost

7.9 XGBoost Parameters: Part A

7.10 XGBoost Parameters: Part B

7.11 Demo: Pima Indians Diabetes

7.12 Model Selection7.13 Common Splitting Strategies

7.14 Demo: Cross Validation

7.15 Key Takeaways





## Lesson 8

### **Recommender Systems :**

8.1 Learning Objectives

8.2 Introduction

8.3 Purposes of Recommender Systems

8.4 Paradigms of Recommender Systems

8.5 Collaborative Filtering: Part A

8.6 Collaborative Filtering: Part B

8.7 Association Rule: Mining

8.8 Association Rule: Mining Market Basket Analysis

8.9 Association Rule: Generation Apriori Algorithm

8.10 Apriori Algorithm Example: Part A

8.11 Apriori Algorithm Example: Part B

8.12 Apriori Algorithm: Rule Selection

8.13 Demo: User Movie Recommendation Model

8.14 Key Takeaways



# CLOUD VISION TECHNOLOGIES

