

Day – 01: Introduction to Big Data and Data Science

- Introduction to Data Science
- Thought Experiment
- Cases of Big Data and Analytics
- Introduction to Cloud Computing
- Artificial Intelligence vs. Machine Learning vs. Deep Learning

Day – 02: Introduction to R

- Basic Programming constructs
- Calculations, Datatypes, Variables, Operators
- Vectors, Lists
- Matrices and Data frame
- Conditional Statements and Loops

Day – 03: Essentials of Statistics for Machine Learning (Part-1)

- Into to stats, Types, basics, Probability
- Random Variable and types (discrete and continuous) – with numeric examples
- Expected Value and spreads - with numeric examples
- Probability distributions (Discrete and Continuous)
- Bayes' Rule
- Central Limit Theorem

Day – 04: Essentials of Statistics for Machine Learning (Part-2)

- Descriptive Statistics (Central Tendency and Dispersion)
- Inferential Statistics
- Confidence Interval
- Hypothesis Testing
- Box Plots, Histograms

Day – 05: Exploratory data analysis in R

- Introduction to Exploratory Data Analysis
- Bi-Variate Analysis
- Scatter Plots, Co-variance and co-relation
- Multi-variate Analysis

Day – 06: Predictive Analytics and Modelling

- Introduction to Predictive Analytics
- Linear Regression
- Multiple Regression
- Logistic Regression
- Industry case

Day – 07: Programming in Python and Machine Learning

- Basic Programming Constructs
- Lists and Dictionaries
- Control Statement, Functions and Modules
- Introduction to Numpy
- Pandas for Data Analysis

Day – 08: Machine Learning (Unsupervised Techniques)

- Introduction and Overview of Machine Learning
- Introduction and Overview of Unsupervised Learning
- Various data issues – sparsity, dimensionality, missing values, distributions, etc.
- Projection based visualization methods e.g. PCA, etc.
- Distance based visualization methods e.g. Multi-dimensional Scaling, etc.
- Partitional Clustering (e.g. k-means, etc.)
- Hierarchical Clustering (e.g. top-down and bottom-up)

Day – 09: Machine Learning (Unsupervised and Supervised Techniques)

Unsupervised Machine Learning

- Fraud detection, Outlier Detection – application of density estimation
- Market Basket Data in Retail and other domains
- Frequent Item-set Mining problem and challenges
- Introduction to MapReduce

Supervised Machine Learning

- Goals of Machine Learning in general
- Various Machine Learning paradigms
- Understanding Data (distributions, visualizations) and Data nuances
- Data transformations (linear, non-linear, normalizations, scaling, etc.)
- Version Spaces and Decision Trees classifiers

Day – 10: Machine Learning (Supervised Techniques)

- K-Nearest Neighbor
- Bayesian classifiers - Naïve Bayes, Various Discriminant Analysis
- Naïve Bayes Text Classifiers
- Support Vector Machines (on Text Classification)
- Bias vs. Variance tradeoff
- Data Pre-processing, Text Data enrichment
- Feature Engineering and incorporating Domain knowledge

Day – 11: Text Analytics (using NLTK)

- **Introduction and Basic Text Processing**
- **Text Classification and Sentiment Analysis**
- **POS Tagging**
- **Parsing Introduction and Dependency Parsing**
- **Information Extraction Named Entity Recognition**

Day – 12: Introduction to Neural Networks and Deep Learning

- **What is deep learning**
- **Motivations for deep architectures and gradient based learning**
- **Stochastic gradient descent**
- **Multi-layered neural networks**
- **Multi-layered perceptron and Applications**
- **Introduction Convolutional neural networks and Recurrent neural networks**