





## FROM ZERO TO HERO

**ONLINE LIVE TRAINING PROGRAM** 

**40 Weeks Program** 

**BATCH 20** 

APRIL

This Batch is Now Accredited





## **Overview**

To maintain a competitive edge, today's organizations must gain practical knowledge of data science, machine learning and artificial intelligence. Careers especially in engineering, data science, and technical management depend on the immense value of those advanced analytical and programming skills.

In the Data Science & Machine Learning program from zero to hero program, an executive education program led by the InnovatiCS, you will dive deep into learning the applied mathematical structure of core machine learning (ML) and artificial intelligence (Al) methods. The course focuses on the skills needed to work efficiently with machine learning models, such classification models, regression models, clustering models, and optimization.

In this course, you will conduct your own analyses and apply the most effective methods to solve various problem types. This is what makes our program unique — we address both the how and the why of machine learning to provide you with the technical knowledge to thrive in any ML-essential field.

You will leave this program with.....

- The tools required to put analytics to practical use and solve specific business problems
- The language and intuition to work effectively with data scientists
- The necessary insights for leveraging analytics to accelerate growth and increase efficiency and productivity



## **Key Benefits from the Program**

Learn from basics of Data Science, R & Python Programming to Advanced ML

Build and Apply: Develop the ability to translate business problems to analytics & ML problems and develop impactful solutions.

Interact with Academic Experts and Industry Practitioners and get mentored - Train on detailed concepts and the nuances of solutioning

Reduce learning curve and be Industry Ready: The in-built Analytics and Al functions of the platform will reduce the learning curve and help to focus more on solutioning rather than just programming.

Dynamic Collaboration: The Collaboration features of the platform will enable to form teams, develop effective solutions and becomebetter team players.

Regular Hackathons: Compete & apply learnings under strict timelines and achieve tangible recognition/personal brand in the Data Science community.

Build and showcase cutting edge ML solutions using the INNOVATICS platform, with long term access to learning content/platform.

Industry standard assessment tests and long-term access to material will enable one's Data Science journey, even after the program completion.



## Who is this program for?



Engineers / Associates / IT
Professionals:
Software engineers in IT/ITES
Startup teams building
ML products/services



Data Analysts/Scientists & Business Analysts: Who want to transition to or progress into data science/ analytical roles and become more efficient and effective in data-driven decision-making.



Managers: Product managers,
Program managers, General
Managers, etc. interested in
improving their analytical skills
and effectively managing
analytics, data science and
machine learning projects



Consultants: Who are driving client projects and looking for acquiring and honing cutting edge analytical and data science skills for a career transition or progress

## Requirements:

- 1-6 Years of work experience
  - Bachelor's /Master's degrees.
- Commitment for Hands On/Lab effort of 8-12 hours per week throughout the program.

#### PREREQUISITES:

This course requires patience, passions, dedication and commitments and we will do our job



## **Key Takeaways**





**Identify** the value in utilizing analytics tools

Gather the skills and confidence necessary to operate in a data-driven environment **Develop** the ability and

intuition to judge "good" analytics" from "bad" analytics" **Understand** the importance of experimentation

platforms to drive business growth **Learn** how to tell a persuasive story

with data visualization tools

**Repeat** unlimited time at no additional cost



## **Program Experience**



Office Hours with Learning Facilitators



**Knowledge Checks** 



Dedicated Program Support Team



Insurance Data



Python/R Coding Exercise in Each Module



**Bite-Sized Learning** 



**Peer Discussion** 



**Faculty Interaction** 



Regular office hours, coaching sessions, and live instructor sessions ensure that you 'get it'.



Graded assignments allow you to get the feedback you need to deepen your understanding of core concepts.



Having a portfolio will allow you to hit the ground running.



## **Program Features**



Live: The program is delivered in an Online format with a highly effective blend of weekend live sessions followed 4 weekdays live sessions, then by self-paced practice sessions spanning across 8 months oriented learning.



**Experiential Learning:** More than 50 industry standard data cases & case studies across domains to be used for applied learning



<u>Collaborative Learning</u>: The INNOVATICS platform enables learners to collaborate on projects, helping them perform better in teams. Develop and showcase a portfolio of projects and solutions



Academic/Industry Leaders Interactions:

100 live sessions planned with Academic and Industry leaders for detailed and application oriented learning.



Milestone based Learning: Entire curriculum is divided into three manageable learning blocks to help the learner reap benefits in an incremental manner. Separate certificate for each block, followed by a certificate for the overall program



<u>Capstone Project:</u> The learner will get the opportunity to participate in live capstone project and develop cutting edge ML solutions under the guidance of experts and implement those, with the opportunity to develop and showcase a portfolio of projects and solutions

## **Sample Case Studies**



Retail/Etail: Market Basket Analysis in Retail -

Application of Association Rule | Mining Technique | Customer Sentiment analysis for an e-commerce retailer | Predict Holiday Sales for A Retail Client | Application of Linear Regression



Human Capital: Application of Non-Hierarchical Clustering in HR Analytics | Domain Recognizing human activity An application of supervised machine learning



Banking & Financial Services: Analyze Credit

Card Spend Data | Application Of Descriptive Analysis

Techniques | Identify risk class and eligibility of a customer |

Application of Machine Learning | Credit Card Attrition



Manufacturing: Estimating Price for Diamonds |
Supervised Learning | Hyper-parameter Optimization |
Predictive maintenance of equipment data | Building predictive



**Healthcare:** Detection of Breast Cancer in A Clinical Trial | Application Of SVM COVID-19 Data Exploration & Visualization



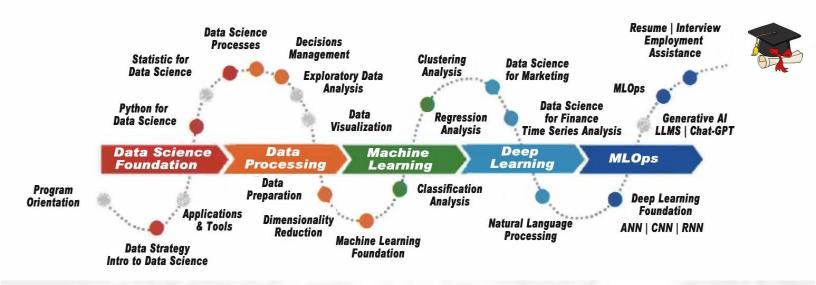
**Sports:** Identify the Top Performing Players in a Domestic Cricket League | Application of Descriptive Analysis Techniques | Fantasy Cricket Team Creation | Application of Linear Programming



## **Learning Journey**

Over the course of eight months, you will develop a competency in data science fundamentals. By working directly with and receiving support from industry-expert mentors, you will be prepared for a shift into a career in data science.

## **Your Portfolio**



#### **Program Highlights**

This curriculum is sourced from 3 Master Degrees (MS-Analytics, MS-Data Science and MS-AI) and 10 years of industry experience. It is organized around the skills that technology giants such as Amazon, Google, Facebook, Apple, and McKinsey value in data science professionals. Numerous participants in the previous batches of the program have leveraged the skills acquired to obtain positions in data analytics oriented roles across a variety of industries.

#### Target Job Functions Include



# Data Science Foundations



## R & Python Programming



# Dedicated Program Support Team



**Insurance Data** 

## **Tools in the Program**



## **Schedule**

#### **Duration**



40 Weeks Online (Live) 12 - 15 hours per week

#### Schedule: 3 Times a week

Sat 12:00pm to 05:00pm UAE Time

Sun 12:00pm to 05:00pm UAE Time

Thur 12:00pm to 03:00pm UAE Time

(Self-Paced)

Office Hourse: Every day per schedule

## Why Now?

#### Why INNOVATICS Data Science Solutions?

INNOVATICShas a reputation for rigor, excellence, and cross-pollination across disciplines. What makes this program dierent?

- ✓ One-to-one career coaching\*
- ✓ Individual feedback
- ✓ Learning in real time
- ✓ Regular live webinars\*
- ✓ Rigorous, graded assignments (Professional-level certificate)
- Assistance with career planning

- Real-world application of knowledge
- ✓ Small group mentoring sessions
- ✓ World-class faculty and thought leadership
- ✓ Learn from and network with your colleagues through peer discussions
- ✓ Ivy League education



<sup>\*</sup>Services provided by Emeritus, a learning partner for this program.

## **Program Faculty**



Dr. Mo Medwani

**Dr. Medwani**, Founder of InnovatiCS (formerly, Big Data Science Solutions), has a PhD in Artificial Intelligence and an expert Data Scientist with a passion for transforming data into useful products. He has over 20 years of experience in Service Delivery Management; Four master's degrees in data science, IT, Machine Learning, and Business Administration; and over 9 years of experience working with Data Science. Mo's specialties include Data Science, Machine Learning, Big Data, Deep Learning, Data Analytics, Application Support and IT Service Delivery Management.



Prof. Ed Bujak
INNOVATICS Co-founder
Python Instructor

**Prof. ED Bujak**, Data science and Python expert withover 26 years of experience in IT and 20 years of education experience in Computer Science, Mathematics, and Engineering. He has two master's degrees in Electrica Engineering and Education, respectively. Ed is a public speak two master's everything related to programming, Eata Ecience, and Eachine Eearning.



Prof. Mohammed Mojeeb INNOVATICS Co-founder Data Engineering Instructor

**Prof. Mohammad Mojeeb,** PBS in Computer science and a Master of Business Adminis-tration (MBA)Microsoft Certified. Technology two master's Specialist in SQL Server Implementation and Maintenance Microsoft Certified. Azure cloud Fundamentals Certified. Computer Programming and Database Development Over 24 years of experience in Software Development, Database Develop-ment, and Cyber Security, and a strong background in all varieties of Application Development and Application Security 10 years of training experience

## **Program Advisors**

Currently working as Manager, Security Business Intelligence at LexisNexis.



Dr. Kirk Borne (Chief Science Officer)

**Dr. Kirk Borne** is the Chief Science Officer at DataPrime, Inc. He is a sought-after global speaker on topics ranginfrom data mining, data management, big data analytics, data science, machine learning, artificial intelligence, the internet of things, data-driven decision-making, modeling & simulation of dynamic systems to emerging technologies, the future of work, education, and science. Kirk provides leadership and mentoring to multi-disciplinary teams of scientists, modelers, and data scientists; and he consults with some external organizations, industries, agencies, and partners in the use of large data repositories and machine learning for discovery, decision support, and innovation. In addition to being a top big data influencer, he was a professor of astrophysics and computational science at George Mason University (GMU) for I2 years. He served as an undergraduate advisor for the GMU data science program and a graduate advisor in the computational science and informatics Ph.D. program.



Salwa Samaoui (InnovatiCS MEA Director)

Salwa Smaoui is a digital transformation leader with over 25 years of experience in the world of technology, marketing and business development, with the majority of her career spent at some of the world's finest global companies. Since February 2016 Salwa has worked as the Microsoft Middle East & Africa Government Business leader. Before that she was the Business Group General Manager for Windows and Surface for Microsoft in Western Europe Region since 2014. Prior to that and in 2011, Mrs Salwa Smaoui was appointed the Director of Advertisement & Online for Middle East and Africa. In this role, Mrs. Smaoui was driving the strategy of Online and Advertisement Microsoft platform to increase its impact and usage by the consumers and businesses in the region. She helped build a strong ecosystem of IT partners, media and advertising partners in the region.



## Program Modules

#### Week 1 Program Orientation

Learning Objectives: The program begins with a lecture on "Why Learning Data Science is an Absolute Must!" where you will be introduced to Data Science process, Data Science portfolio, analytics types, day to day activities, quantitative & statistical techniques required & the career path to start the journey.

	Program Orientation (Agenda – Curriculum)
	<ul> <li>Program Orientation (Agenda – Curriculum)</li> <li>InnovatiCS Website   Canvas   Slack walk through</li> </ul>
Session 2	Why should you become a Data Scientist?
	<ul> <li>Data Explosion</li> <li>Why Data Science?   What is Data Science?   Type of Analytics</li> <li>Data Science Portfolio   Data Science Process   Career in Data Science</li> </ul>
Session 3	Introduction to Data Science   CAP Certificate
	<ul> <li>Introduction to Data Science</li> <li>Introduction to Machine Learning</li> <li>Introduction to Deep Learning overview</li> <li>Certified Analytics Professional (CAP)</li> </ul>
Session 4	Projects, Teams & Team Leads
	<ul> <li>Introduction to Data Science Project</li> <li>Projects Discussion (Milestone Projects Assignment)</li> <li>Projects List   Project Templates   Project Team</li> </ul>
Session 5	Program Tools & Installation 2
	<ul> <li>Jupyter Notebook</li> <li>MySQL: Installation</li> <li>Excel</li> <li>Tableau: Installation</li> <li>RapidMiner: Installation</li> </ul>

#### Week 2 Business Understanding (Strategy | Literacy | Problem Framing)

Learning Objectives: The basic workflow is now in place. You will dive into the first stage of the Cross -Industry Standard Process for Data Mining (CRISP-DM) process which deals with understanding analytical project objectives from a business perspective. Customers often have competing objectives and constraints that must be properly identified and balanced. The goal is to uncover important factors that could influence the outcome of the analytical projects. Neglecting thisstep can mean that a great deal of effort is put into producing the right answers to the wrong questions. In this section also we will provide an overview of the importance of data strategy as a tool that enables any organization to make better-informed decisions, improve business processes and gain new revenue streams. Data literacy involves articulating a problem that can potentially be solved using data.

# Session 1 Introduction to Data Literacy and Data Strategy Deciding Your Strategic Data Needs Using Data to Improve Your Decisions | Using Data to Understand Your Customers and Markets | Using Data to Provide More Intelligent Services | Using Data to Make More Intelligent Products | Using Data to Improve Your Business Processes Monetizing Your Data



Session 2	<ul> <li>Defining Your Data Use Cases   Sourcing and Collecting the Data   Data Governance  </li></ul>
450	<ul> <li>Deep Learning Explained</li> <li>Biological Neural Network</li> </ul>
Session 3	Introduction to Big Data
	<ul> <li>The job market around Big Data</li> <li>What is Big Data?</li> <li>Big Data use cases &amp; Big Data Ecosystem</li> <li>HDFS   What is MapReduce   Pig, Hive etc.</li> <li>NoSQL Databases</li> </ul>
Session 4	Introduction to CRISP-DM
	<ul> <li>Solutions Methodologies (Macro vs. Micro)         <ul> <li>Scientific Research Method</li> <li>Operations Research Method   Water Fall Method   CRISP-DM</li> <li>Exploration &amp; Discovery</li> <li>Solutions are Dependent on Data   Solutions are Independent on Data</li> </ul> </li> <li>CRISP-DM Process</li> <li>Business Understanding   Data Understanding   Data Preparation   Modeling   Evaluation   Deployment</li> </ul>
Session 5	Decisions Management & Problem Framing
	<ul> <li>Quantifying Business Problem</li> <li>Converting Business Problems into Analytics Solutions</li> <li>Defining the objectives, assumptions, contains, analytical approaches.</li> <li>Designing the Analytics Base Table &amp; Implementing Features</li> <li>Identifying Y variable &amp; Measures of Success</li> <li>Stakeholders &amp; Analytical team</li> </ul>
Week 3	Data Science Foundations (Python Programming   GitHub)
Students will work Students will learn	s: In this Week, students will be introduced Python programming language commonly used with the tools that professional Data Scientists work with, like Jupyter Notebooks, IDE, and others about what each tool is used for, what codes they can execute, and their features and limitations
Session 1	Basic Python Programming 1
60.8	<ul><li>Introduction to Python &amp; Jupyter Notebook</li><li>Basic operators in Python</li></ul>



Data Structure: Sting | Lists

Session 2	Basic Python Programming 2
	Data Structure
	o Tuples   Sets   Dictionaries
Session 3	Intermediate Python Programming
	Python data types (basic and Boolean)
A 17.50	Conditional Statements.
	<ul> <li>Python built-in data types</li> </ul>
	Basic operators in Python
Samian 4	Loop and control statements like break, if, for, continue, else, range () and more
Session 4	Advanced Python Programming
	Functions, Assignment & Operations
	Conditions & Loops
Session 5	Introduction - Git and GitHub
-32-37	Introduction To Git
	Installing Git   Configure Git
	Basic commands
	Exploring Git log   Git diff
	Making corrections Git reset
	Branching
Week 4	Data Science Foundations (Statistics - Probability - SQL)

Learning Objectives: In this Week, students will build a solid foundation of statistics for Data Science, learn probability, distributions, hypothesis testing and SQL. Statistics is the science of assigning a probability to an event based on experiments. It is the application of quantitative principles to the collection, analysis, and presentation of numerical data. Ace the fundamentals of Data Science, statistics, and Machine Learning with this course, it will enable students to define statistics and essential terms related to it, explain measures of central tendency and dispersion, and comprehend skewness, correlation, regression, distribution. Students will be able to make data-driven predictions through statistical inference. We will also learn about creating and working with tables, Databases Data Definition Language (DDL) and Data Manipulation Language (DML), Data Types SELECT, INSERT, UPDATE, AND DELETE Statements in SQL.

Session 1	Introduction to Statistics
	Understanding Data
	Levels of Measurement
	Measures of Central Tendency
	Population and Sample
	Measures of Dispersion
	Quartiles and IQR
Session 2	Advanced Statistics / Probability & Information Theory
	Introduction to Distribution
	Uniform Distribution   Binomial Distribution
	Poisson Distribution   Normal Distribution
	Skewness   Standardization and Z Score
	Central Limit Theorem
0110 120	Hypothesis Testing
(d) A - 1 - 1 - 1	ANOVA - Analysis of Variance
	Chi-Square Analysis



<b>Session 3</b>	Crash Course in SQL (Tables and Constraints)
	Creating, and Working with Tables
	Default Constraints
	Cascading referential integrity
	Check Constraint   Identity column   Unique key constraint.
	Get last generated id in SQL server.
Session 4	<b>SQL Statements Group by Clause</b>
	SQL Statements (Select, Insert, Delete, Update)
	Group by Clause
	<ul> <li>Difference between where and having in SQL server.</li> </ul>
	Basic Joins   Advance Joins   Self Joins
	Different ways to replace NULL   Coalesce function.
	Union and Union All
Session 5	Team Projects 1st Presentation (Framing Business Problem)
	Team Projects (Business Understanding)
	Team Leads' First Presentation
	Team Leads 1 list i resentation
dents learn conce marily on someth	Data Science Foundations (Visual Data Sensemaking)  Before you can present information to others, you must understand its story. In this Week pts, principles, and practices of visual data sense-making. The skills taught in this book rely ing that most of us possess—vision—interactively using graphs to find and examine the and relationships that reside in quantitative data.  Building a framework for Visual Data Sensemaking
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#### Week 6

#### **Data Understanding | EDA & Statistical Analysis**

Learning Objectives: In the Week, students will be introduced the second phase of the Cross- Industry Standard Process for Data Mining (CRISP-DM) process model. Students obtain data and verify that it is appropriate for the needs. Students learn to identify issues that cause the analyst to return to the 'Business Understanding' phase of the project and revise the plan. Students may even discover flaws in the 'Business Understanding', another reason to rethink goals and plans. The Data Understanding phase includes four tasks. These are Gathering data — Describing data — Exploring data —Verifying data quality.

Describing data – E	Exploring data – Verifying data quality.
Session 1	Data Loading & Manipulation
	<ul> <li>Read data from different sources.</li> <li>Identifying Categorical Data: Nominal, Ordinal and Continuous</li> <li>Univariate   Bivariate   Multivariate Analysis</li> <li>Types of Data   Type of Central Tendency   Data distribution   Data Density</li> </ul>
Session 2	<ul> <li>Data Types- Measure of Shape - Position - Dispersion</li> <li>Measures of the Spread: Range – IQR – Variance – Standard Deviation</li> <li>Measures of Dispersion</li> <li>Measures of Position: Statistical Analysis (John T. 5 Numbers Summary)</li> <li>Measures of Relationships: Correlation   Variance &amp; Covariance</li> <li>Measures of Shapes: Skewness &amp; Kurtosis</li> </ul>
Session 3	<ul> <li>Data Visualization (Numerical Data &amp; Graphical Descriptive Statistics)</li> <li>Data Visualization (Histograms, Bar Plot, Scatter Plot, Box Plot)</li> <li>Data Quality Report   Develop the Code Book         <ul> <li>O Summary of Data   Type of variable   Ranges of variables   Missing fields   Identify the primary list of variables to solve the business problem</li> </ul> </li> <li>Detecting missing values &amp; outliers   Duplicates &amp; redundant records</li> </ul>
Session 4	NumPy Array   The Shape and Reshaping of NumPy Array
	<ul> <li>NumPy ndarray Data Type, Features, and Array Dtype</li> <li>Matrices, Dimensions, Axis, Shape, Reshape</li> <li>Sequences (Index and Slice), Iterable</li> </ul>
Session 5	ML with RapidMiner   Tableau   SAP Analytics Cloud
	<ul> <li>RapidMiner         <ul> <li>Installation   RapidMiner Basics   Studio - GUI Intro   Visualizing Data in RapidMiner Marketplace – Extensions</li> </ul> </li> <li>SAC         <ul> <li>Using in-story calculations   Add a measure input control   Conditional formatting  </li></ul></li></ul>



#### Week 7

#### **Data Preparation & Preprocessing**

Learning Objectives: In the Data Preparation phase, students learn the process of cleaning and transforming raw data prior to processing and analysis. This is an important step prior to processing and often involves reformattingdata, making corrections to data and the combining of data sets to enrich data. Students also learn ho

ow to fix data quali	ty issues discovered through EDA & Data visualization from the DataUnderstanding phase
Session 1	Data Preparation & Processing (Categorical Attributes)
	<ul> <li>Encoding Categorical Data</li> <li>Replacing values   Creating Dummy Variables   Encoding labels   One-Hot encoding   Binary encoding   Backward difference encoding   Miscellaneous Features</li> </ul>
	<ul> <li>Data Quality Report</li> <li>Develop the Code Book         <ul> <li>Summary of Data   Type of variable   Ranges of variables   Missing fields   Identify the primary list of variables to solve the business problem.</li> </ul> </li> </ul>
Session 2	Data Preparation & Processing (Numeric Attributes & Transformation)
	<ul> <li>Statistical Analysis (John T. 5 Numbers Summary)</li> <li>Variance &amp; Covariance</li> <li>Correlation Analysis</li> </ul>
Session 3	Data Preparation & Processing (Missing Values   Outliers   Duplicates)

- Dealing with Skewness & Kurtosis | Missing Values & Outliers | Duplicates & redundant
- Variable Conversion
  - o Discretization | Binning
- Variable Transformation
  - o Normalization | Standardization | Factorizations | Binarization
- **Box-Cox Transformation**
- **Data Partitioning**

#### **Session 4**

#### Expanding, Squeezing, Indexing and Slicing of NumPy Array

- Filtering, Sorting, Statistics
- Correlation, Regression, Outliers
- **Advanced Operations**

#### Session 5

#### Data Prep with RapidMiner | Excel

#### RapidMiner:

- o Turbo Prep Introduction | Data Cleansing | Merging Data
- o Data Pivoting | Data Preparation | Connecting to Databases

#### **Excel**

- o Inserting a Line Break with Alt + Enter | Create Easily Printable Excel Documents
- o Insert Hyperlinks into Excel Spreadsheets | Using Excel's Freeze Panes to Handle Large Datasets | Introduction to Excel's Pivot Tables
- o Macros | Absolute and Relative Cell References
- o Create Dynamic Names in Excel Spreadsheets
- o Using Named Ranges to Make Formulas More Readable

#### Tableau:

- Time series, Aggregation, and Filters | Working with Data Extracts in Tableau
- Working with Time Series | Understanding Aggregation, Granularity, and Level of Detail | Creating an Area Chart & Learning About Highlighting | Adding a Filter and **Ouick Filter**



<ul> <li>Local Linear Embedding (LLE)         <ul> <li>Intuition   Least squares problem</li> </ul> </li> <li>Eigenvalue problem         <ul> <li>Data Prep with RapidMiner   Excel</li> </ul> </li> <li>RapidMiner:             <ul> <li>Data Cleansing   Merging Data</li> <li>Data Pivoting   Data Preparation   Connecting to Databases</li> </ul> </li> <li>Excel             <ul> <li>Add a Drop-down List in Excel   Sort Multiple Columns Within a Table</li> <li>Saving Time in Excel and Doing Everything Faster by Using Excel Shortcuts</li> <li>Find and Replace – References   Find and Replace - Formatting</li> <li>Beauty Saving - The Professional Way of Saving Files</li> <li>The Power of F2   Conditional Formatting</li></ul></li></ul>
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o Intuition   Least squares problem
Local Linear Embedding (LLE)
Introduction to Manifold Learning (t-SNE, LLE, Isomap)
ISOMAP   Locally Linear   Modified Locally Linear Embedding
Aggregation   Sampling   Dimensionality Reduction
Factor Analysis: LDA & SVD
Singular Value Decomposition & Linear Discriminant Analysis
Discretization & Binarization   Variable Transformation   Data Partitioning
Feature Subset Selection   Feature Creation
Aggregation   Sampling   Dimensionality Reduction
Factor Analysis: PCA   KPCA
Feature Subset Selection   Feature Creation
Regularization & Embedded Feature Selection
Machine Learning Method   Feature Importance
Correlation Method
<ul> <li>Goals of Feature Selection</li> <li>Classes of Feature Selection Methodologies</li> </ul>
Principal Component Analysis     Goals of Feature Selection     Classes of Feature Selection Methodologies

Factor Analysis & Dimensionality Reduction



Week 8

#### Week 9

#### **Supervised Learning - Classification**

Learning Objectives: This week broadens concepts learned in Data Understanding, Data Preparation & Exploratory Data Analysis (EDA) by extending to Machine Learning where you will learn about the models &methods used in machine learning & apply them to a real-world. The aim of supervised machine learning is to build a model that makes predictions based on evidence in the presence of uncertainty. In this session, you will learn about the different algorithms of supervised learning such as Decisions Tree, Rule-Based classifier & Naive Bayes, Logistic Regression, Support Vector Machine, Nearest Neighbor Classifiers with deeper use of scikit-learn functionality, introducing automated methods of feature selection, options for estimation including stochastic gradient descent, & advanced metrics for model evaluation.

ncluding stochasti	c gradient descent, & advanced metrics for model evaluation.
Session 1	<b>Decision Tree &amp; Rule-Based Classification</b>
	Algorithm for Decision Tree Induction
	Methods for Expressing Attribute Test Conditions
	Measures for Selecting an Attribute Test Condition
	Characteristics of Decision Tree Classifiers
	How a Rule-Based Classifier Works
	Characteristics of Rule-Based Classifiers
Session 2	Naïve Bayes & Logistic Regression
	Naïve Bayes motivation
	Naïve Bayes Mathematical foundation
	Basics of Probability Theory
	Naïve Bayes Assumption
	Logistic Regression motivation   Mathematical foundation
	Logistic Regression as a Generalized Linear Model
	Characteristics of Logistic Regression
Session 3	Support Vector Machine SVM & KSVM
	SVM motivation
	SVM Mathematical foundation
	Margin of a Separating Hyperplane
	Linear SVM   Nonlinear SVM
	Soft-margin SVM
	Characteristics of SVM
Session 4	Neighbor Classifiers (KNN)
	KNN motivation
	KNN Mathematical foundation
	KNN Algorithm
	Characteristics of Nearest Neighbor Classifiers
Session 5	Pandas Fundamentals
	Pandas Data Frame Basic
	Pandas' data types: Index, Series, DataFrame   Axis
	Operations and Methods on Panda's data types
	Create Data Frame   Addition and Deletion   Contact and Merge
	<ul> <li>Export to CSV   Excel   Txt   Group By   Sorting   Stack &amp; Unstack   Iterate Over Data Frame</li> </ul>



#### Week 10

#### **Supervised Learning - Regression**

Learning Objectives: This week provides Regression Analysis (Supervised Learning with a continuous target variable). Regression analysis is a form of predictive modeling technique that investigates the relationship between a dependent (target) and independent variable (s) (predictor). This technique is used for forecasting, timeseries modeling & finding the causal effect relationship between the variables. Machine learning topics taught involve linear regression algorithms (Simple Linear Regression, Multiple LinearRegression & Polynomial Linear Regression, Decision Tree, and Support Vector Machine) and Regularization Techniques (Lasso, Ridge, Elastic Net, Least Squared) with deeper use of scikit-learn functionality.

Session 1	Simple Linear Regression & Polynomial Regression
	<ul> <li>Preparing Data for Linear Regression</li> <li>Making Predictions with Linear Regression</li> </ul>
	Polynomial Algorithm
No.	Why use polynomial regression?
	Polynomial Features
Session 2	Multiple Linear Regression
	Multiple Linear Regression
	• The F-Statistic
	Interpreting results of Categorical variables
	Heteroscedasticity
	Backward Elimination   Backward Elimination   Automatic Backward Elimination
Session 3	Least Square/Lasso/Ridge/Elastic Net Regression
	OLS Regression- Theory   Implementation
	Confidence Interval and OLS Regressions
	Ridge Regression   LASSO Regression
	Implement ANOVA on OLS Regression
	Identify Multicollinearity   Partial Least Square Regression
Session 4	Decision Tree Regression & Support Vector Machine SVR
	A Basic Algorithm to Build a Decision Tree
	<ul> <li>Methods for Expressing Attribute Test Conditions   Measures for Selecting an Attribute Test Condition</li> </ul>
	Algorithm for Decision Tree Induction   Characteristics of Decision Tree Regression
	Margin of a Separating Hyperplane   Linear SVM   Soft-margin SVM   Nonlinear SVM
	Characteristics of SVM
Session 5	Using Pandas for Data Analysis & Data Wrangling 1
	Data types: Categorical, Continuous
	EDA, Data Cleaning
	Filtering, Row/Record/Column
Week 11	Ensemble Methods (Bagging & Boosting & Stacking)

Learning Objectives: Ensemble methods help to improve the predictive performance of Machine Learning models. In this Week, you will learn about different Ensemble methods that combine several Machine Learning techniques into one predictive model to decrease variance, bias or improve predictions. You will learn about selecting one model over another and boosting & its importance in Machine Learning. You willlearn how to convert weaker algorithms into stronger ones.



Session 1	Random Forest Voting & Averaging
S 2	<ul> <li>Random Forests</li> <li>Empirical Comparison among Ensemble Methods</li> </ul>
Session 2	<ul> <li>Bagging   Boosting   Gradient Boosting   Ada-Boost   Stacking</li> <li>Methods for Constructing an Ensemble Classifier</li> <li>What is the Model Selection?   The need for Model Selection</li> <li>Cross-Validation</li> <li>What is Boosting?   Adaptive Boosting</li> <li>How Boosting Algorithms work?</li> <li>Types of Boosting Algorithms</li> </ul>
Session 3	Class Imbalanced Problem
	<ul> <li>Building Classifiers - Class Imbalance</li> <li>Data-based Approaches &amp; Algorithmic approach   Evaluating Performance - Class Imbalance</li> <li>Finding an Optimal Score Threshold   Aggregate Evaluation of Performance</li> </ul>
Session 4	Using Pandas for Data Analysis & Data Wrangling 1
	<ul> <li>Column/Feature Manipulations</li> <li>Statistical Foundations</li> <li>Problematic Data - Outliers, Missing Data (Impute)</li> </ul>
Session 5	ML with RapidMiner   Excel
	<ul> <li>RapidMiner:         <ul> <li>Applying the Model   Testing a Model   Validating a Model   Finding the Right Model  </li> <li>Optimization of the Model Parameters</li> </ul> </li> <li>Excel</li> </ul>
	<ul> <li>Ordering the Source Worksheet</li> <li>Using Lookup Functions (Vlookup) to Fill the Database Sheet</li> </ul>
	<ul> <li>Using Lookup Functions (Vlookup) to Fill the Database Sheet</li> <li>Use Sumif to Complete the Database Sheet</li> </ul>
	<ul> <li>Using Index &amp; Match as a Substitute for Vlookup</li> </ul>
	<ul> <li>Learn How to Find Mistakes with Countif</li> </ul>
	<ul> <li>Calculating Growth Rates in Excel</li> </ul>
	• Tableau:
	<ul> <li>Joining, Blending and Relationships   Dual Axis Charts   Joins with Duplicate Values  </li> <li>Joining on Multiple Fields   The Showdown: Joining Data vs. Blending Data in Tableau</li> </ul>
	<ul> <li>Creating Calculated Fields in a Blend (Advanced Topic)   Working with Relationships in Tableau</li> </ul>
Week 12	Model Evaluation & Optimization

<u>Learning Objectives</u>: Model building is an iterative process. Employing Feature Engineering techniques along witha careful model selection exercise helps to improve the model. Further, tuning the model is an important step to arrive at the best possible result.

This Week talks about the steps and processes around the same, you will learn how to analyze the performance of each algorithm, and dive deep in core Machine Learning concepts like Cost Function, Object Function, Model Optimization, Model Tuning, Regularization, Gradient Boosting, Grid & Random Search.



Session 1	Model generalization: Assessing Predictive Accuracy for New Data
	CM, ROC, Rank-Ordered Approach
	R2, MSE, MAE, Median Error, Median Absolute error, Correlation      Research for Market Over 6th in the Control of the Co
426	<ul> <li>Reasons for Model Overfitting</li> <li>Model Selection   Using a Validation Set</li> </ul>
100	Incorporating Model Complexity   Estimating Statistical Bounds
1900	Model Selection for Decision Trees
Session 2	<b>Evaluation of Classification Models &amp; Regression Models</b>
	Holdout Method & Cross-Validation
	Presence of Hyper-parameters & Hyper-parameter Selection
	Nested Cross-Validation
San and	Pitfalls of Model Selection & Evaluation
	Overlap between Training & Test Sets
	Use of Validation Error as Generalization Error   Cluster Evaluation
Session 3	Model Optimization through Parameter Tuning   XGBoost
	Estimating the Confidence Interval for Accuracy
	Comparing the Performance of Two Models
~	• XG-Boost
Session 4	Data Wrangling & Visualizing with Pandas
	Data transformations, Data type, Scaling, Normalizing, Binning, Group by
	• Slicing, changing index, Data conversion, Joining and Merging, Concatenation, Columns,
	Pivoting, Melting, Handling Duplicates
G • F	Distributions, Histograms, Box Plot, Line Plots, Bar Plots, Grouped Bar Plot, Scatter Plot
Session 5	Team Projects 3rd Presentation (Data Preparation)
	Team Projects (Analytical Approach)
	Team Leads' Third Presentation
Week 13	Unsupervised Learning - Clustering Analysis
	: In this week, we will shift to Unsupervised Learning techniques (target variable is unknown).  ning finds hidden patterns or intrinsic structures in data. The end goal is less clear-cut than

Session 1	K-Mean Clustering
	The Basic K-means Algorithm
	Bisecting K-means
	K-means & Different Types of Clusters
	K-means as an Optimization Problem
Session 2	Agglomerative Hierarchical Clustering
	Basic Agglomerative Hierarchical Clustering Algorithm
NOW A	AHC Specific Techniques
100	The Lance-Williams Formula for Cluster Proximity
Carlo Company	Key Issues in Hierarchical Clustering
	AHC Outliers



Session 3	DBSCAN   Mean Shift   Gaussian Mixed Models   Fuzzy C Means
	BIRCH   DBSCAN   Mini-Batch K-Means
	Mean Shift   OPTICS  Spectral Clustering  Gaussian Mixture Model
Session 4	Python Regular Expression
100	Extracting data from the web using JSON, Google API, and XML
	Data Extraction - Getting Data from the Internet
Session 5	ML with RapidMiner   Excel
	RapidMiner:
	<ul> <li>Logistic Regression   Times Series   Clusters   Anomaly detection   AssociationDiscoveries</li> </ul>
	• Excel
	<ul> <li>Introduction to Excel Charts</li> </ul>
	<ul> <li>Modifying Excel Charts - The Easy Way</li> </ul>
	<ul> <li>Creating a Bridge Chart in Excel</li> </ul>
	<ul> <li>Stacked Column Chart with a Secondary Axis</li> </ul>
	<ul> <li>Doughnut Chart   Area Chart   Bridge Chart</li> </ul>
	• Tableau:
	<ul> <li>Table Calculations, Advanced Dashboards   Storytelling</li> </ul>
	<ul> <li>Mapping: How to Set Geographical Roles   Creating Table Calculations for Genderal</li> </ul>
	<ul> <li>Creating Bins and Distributions for Age   Leveraging the Power of Parameters</li> </ul>
	<ul> <li>How to Create a Tree Map Chart   Creating a Customer Segmentation Dashboard</li> </ul>
	<ul> <li>Advanced Dashboard Interactivity   Analyzing the Customer Segmentation Dashboard</li> </ul>
Week 14	Anomalies Detection & A/B Testing

Learning Objectives: This Week is a continuation of the Unsupervised Machine Learning algorithms from the previous week. We go deep into Association Rule mining which has numerous applications that are widely used to help discover sales correlations in transactional data or in medical data sets. We also go deep to A | B Testing, also known as split tests, to cover the ins & outs of how to use Python & R toanalyze customer behavior & business trends as well as how to create, run, and analyze A | B tests to make proactive, data-driven business decisions.

Session 1	Classification Based Algorithms - KNN
	Characteristics of Anomaly Detection Problems   Anomaly Detection Methods
	Spectrum of Anomaly detection techniques
	Nearest Neighbor based Methods   Rule-based Anomaly Detection
	Classification based Models for Anomaly Detection
	Proximity-based Approaches
Session 2	Nearest-Neighbor Based algorithms
	Nearest-Neighbor based algorithms.
	o k-NN Global Anomaly Score   Local Outlier Factor (LOF)
	<ul> <li>Connectivity based Outlier Factor (COF)</li> </ul>
- OF 6	<ul> <li>Local Outlier Probability (LoOP)</li> </ul>
	<ul> <li>Influenced Outlierness (INFLO)  Local Correlation Integral (LOCI)</li> </ul>
W 188	Clustering based algorithms.
1-1-1	<ul> <li>Cluster based Local Outlier Factor (CBLOF)</li> </ul>
	<ul> <li>Local Density Cluster based Outlier Factor (LDCOF)</li> </ul>



Session 3	Clustering Based algorithms
	<ul> <li>Clustering based algorithms.</li> <li>Cluster based Local Outlier Factor (CBLOF)</li> <li>Local Density Cluster based Outlier Factor (LDCOF)</li> </ul>
Session 4	A/B Testing
	<ul> <li>Overview of A   B Testing?</li> <li>How A   B Testing Works</li> <li>Policy and Ethics</li> <li>Characterizing Metrics</li> <li>Designing an Experiment</li> <li>A   B Testing Process   Analyzing Results   Alternatives to A   B Testing</li> </ul>
Session 5	Python: BeautifulSoup Library
	<ul> <li>Installing Beautiful Soup</li> <li>Data extraction with BeautifulSoup</li> <li>BeautifulSoup Usage   Filtering   BeautifulSoup Object</li> <li>Extracting all the URLs found within a page 'a' tag   Extracting text from a page.</li> </ul>
Week 15	Association Rules   Recommender Systems

Learning Objectives: This Week is a continuation of the Unsupervised Machine Learning algorithms from the previous week. We go deep into Association Rule mining which has numerous applications that are widely used to help discover sales correlations in transactional data or in medical data sets. We will also dive deep into recommender systems, which are software that select products to recommend to individual customers. You will learn how to produce successful recommender systems that use past product purchase and satisfaction data to make high-quality personalized recommendations.

Session 1	Association Rules - Apriori	
	Strength of an association rule	
	The Apriori Principle	
	Frequent Itemset Generation in the Apriori Algorithm	
	Candidate Generation & Pruning	
	Computational Complexity	
Session 2	Association Rules - Eclat	
	Horizontal vs Vertical Data Format	
	The Intuition of ECLAT Algorithm	
	Advantages of Eclat & Eclat vs Apriori	
Session 3	Recommender Systems: First Generation	
	Matrix Factorization Model (Object recommendation)	
	Content Filtering   Collaborating Filtering (CF)	
<b>Session 4</b>	<b>Recommender Systems: Second Generation</b>	
1.05	CF Neighborhood-Based Approach	
	CF Location-Based Approach	
	Popularity Based   Content Based	
Session 5	Recommender Systems: Third Generation	
	Collaborative filtering	
	Matrix Factorization (MF)	



#### Week 16

#### **Time Series Analysis**

**Learning Objectives:** In this Week, you will learn about Time Series Analysis to forecast dependent variables based on time. Time series analysis is a statistical technique that deals with time-series data, or trend analysis. Timeseries data means that data is in a series of periods or intervals. You will learn different models for time series modeling such that you analyze a real time-dependent data forforecasting.

Session 1	AR - MA & EST Models
A 400	What is Time Series Analysis?
	Importance of TSA   Components of TSA
	Single Exponential Smoothing
	Forecasting with Single Exponential Smoothing
	Double Exponential Smoothing   Forecasting with Double Exponential Smoothing
The second	Triple Exponential Smoothing
Session 2	ARMA & ARIMA Models
	• Forecasting
	Relation between time series: Causality & time lags
	Distinction between short & long run
	Study of agent's expectations
	Stationarity   ACF & PACF
	Trend removal   Seasonal adjustment
Session 3	SARIMAX Model
	The limitations of ARIMA
	The SARIMA extension of ARIMA
	<ul> <li>Implementing SARIMA method using the Statmodels library</li> </ul>
Session 4	Python: Matplotlib Library
	• Annotation   Autocorrelation   Curves   Scales   Shapes   Box Plot   Violet Plot   Histograms   Pie Chart   Twin Axis   Figures   Color Map Color Palettes/Figure Aesthetics
Session 5	Team Projects 4th Presentation (Analytical Approach Implementation)
	Team Projects (Analytical Approach Implementation)
	Team Leads' Fourth Presentation
Week 17	Text Analysis & Natural Language Processing (NLP)

Learning Objectives: In this Week, we dive into Natural Language Processing or NLP which is one such technology penetrating deeply and widely in the market, irrespective of the industry and domains. It is extensively applied in businesses today and it is the buzzword in every engineer's life.

#### **Session 1**

#### **Introduction to Natural Language Processing**

- What is Natural Language Processing?
- History of Natural Language Processing
- NLP Applications | Levels | Components
- NLU vs NLG
- NLP Pipeline and Tasks | NLP Toolkits and Libraries
- NLP Challenges



Session 2	Essentials of Natural Language Processing
	Basic Text Analysis
	Tokenization   POS Tagging
	Stop Word Removal
	Text Normalization
6.55	Spelling Correction
	Stemming   Lemmatization
for the	Named Entity Recognition (NER)
	Word Sense Disambiguation
	Sentence Boundary Detection
Session 3	NLP Feature Extractions
	Data Structures
100	NLP Pre Processing
	The Bag of Words
	Frequency Vector (Count of Vectorization)
	One-Hot Encoding
200	Term Frequency – Inverse Term Frequency
	Distributed Representation
	Word Embedding
Session 4	NLP with TextBlob & spaCy
	Introduction to TextBlob
	Language Detection
	POS   Word Inflection
	Sentiment Analysis
	Introduction to spaCy Library   Objects of spaCy Library
	The Statistical Modeling   Processing Pipelines
Session 5	Python: SciPy Library
	Introduction to SciPy, building on top of NumPy.
	What are the characteristics of SciPy?
	Various sub packages for SciPy like Signal, Integrate, Fatback, Cluster, Optimize, Stats and
	more, Bayes Theorem with SciPy.
Week 18	Advanced Natural Language Processing (NLP)
	In this Week, we dive into advanced Natural Language Processing which is one such g deeply and widely in the market, irrespective of the industry and domains. It is extensively
	today and it is the buzzword in every engineer's life.
Session 1	Text Classification
	Data preparation   Data Vectorization
	Text Classification
	Text Classification
Session 2	Text Summarization
	What is Text Summarization?
1111 1111	Text Summarization Categories
VII A TOTAL	Stages of Text Summarization



Session 3	Topic Modeling
	What is Topic Modeling?
	Topic Modeling Use cases
3357	Topic Modeling Libraries
	Latent Semantic Analysis (LSA)
0.00	Latent Dirichlet Allocation (LDA)
	Hierarchical Dirichlet Process (HDP)
Session 4	Sentiment Analysis
	What is Sentiment Analysis?
	Types of Sentiment Analysis
	Benefits of Sentiment Analysis
	Examples of Sentiment Analysis
	Challenges of Sentiment Analysis
Session 5	Large Language Models (LLMs) - ChatGPT, GPT4, BERT
	• What are LLMs? Why LLMs (such as GPT and Bard) is so special?
	The Power of Attention Mechanisms   Encoder-Decoder Architectures
	Demystifying Transformers   Learn and Understand Transformers
	Scaled Dot Product Attention   Multi-Headed Attention
	Transfer Learning Potential
	Exploring the GPT Architecture: Foundations   Innovations
	<ul> <li>Practical Applications of Masked Multi-Head Attention</li> </ul>
	GPT Pre-Training Strategies and Techniques
	Optimizing GPT Pre-Training for Performance
Week 19	Deep Learning: ANN   CNN   Object Detection
	: This week, we dive into Deep Learning - an Artificial Intelligence function that imitatesthe nan brain in processing data and creating patterns for use in decision making. Deep Learning

Networks. You'll learn how Deep Learning Networks can be successfully applied to data for knowledge discovery knowledge application, and knowledge based prediction.

Session 1	Introduction to Deep Learning
	Deep Learning Explained
	Biological Neural Network
	Perceptron – Gradient Decent – Scholastic Gradient Descent
Session 2	Artificial Neural Network (ANN)
	ANN Explained
	Multi-layer Neural Network
	Characteristics of ANN
Session 3	Convolutional Neural Network (CNN)
	CNN Explained
	Synergistic Loss Functions   Responsive Activation Functions
	Regularization   Initialization of Model Parameters
	Characteristics of Deep Learning



<b>Session 4</b>	Real Time Object Detection using YOLO V4
	Introduction to YOLO
	YOLO as a real-time object detector.
	YOLO as an object detector in TensorFlow & Keras
	How to train your custom YOLO object detection model
	YOLO compared to other detectors.
Session 5	Python: Torch Library 1
	How to Install PyTorch
	PyTorch Deep Learning Model Lifecycle
	o Prepare the Data
	o Define the Model
	o Train the Model
	o Evaluate the Model
	o Predictions.
Wools 20	Solf Organizing Mana   Poltzmann Machines   Automoders

Week 20 Self-Organizing Maps | Boltzmann Machines | Autoencoders

Learning Objectives: This week is a continuation of week 19. You will learn 3 more Deep Learning Models (SOMs, BM and AutoEncoders). In SOMs, you will learn How to return the specific features (like frauds) detected by the SOM and How to make a Hybrid Deep Learning Model. You will also learn the Energy-Based Model point of view, and then for the Practical Lectures we will focus more on the Probabilistic Graphical Model point of view. Finally, you will learn how to build an AutoEncoder from scratch with PyTorch and how to manipulate classes and objects to improve and tune your AutoEncoder.

Session 1	Introduction to SOMs, BMN & AutoEncoders
	Introduction to Self-Organizing Maps
	Introduction to Boltzmann Machine
	Introduction to Auto Encoder
Session 2	Self-Organizing Maps
	How Self-Organizing Maps work?
	How Self-Organizing Maps Learn?
	Live SOM example
	• Energy-based Models (EBMs)
	Restricted Boltzmann Machines (RBM)
	Deep Boltzmann Machines (DBM)
Session 3	<b>Boltzmann Machines</b>
	Energy-Based Models (EBM)
	Deep Belief Networks
	Deep Boltzmann Machine
Session 4	AutoEncoders
	Training an Auto Encoder
F 6/1	Overcomplete hidden layers
15-70F-J	Sparse Autoencoders   Denoising Autoencoders   Contractive Autoencoders
	Stacked Autoencoders   Deep Autoencoders
Session 5	Team Projects Final Presentation (Insights & Action Plan)
15	Team Projects (Analytical Approach Implementation)
	Team Leads' Final Presentation



#### Week 21

#### **Sequence Learning & GANs**

Learning Objectives: This Week talks about yet another interesting implementation of Neural Networks that revolves around equipping computers to understand human language. You will learn to work with text data and sequential data, and explore the interesting world of RNNs and LSTMs

Session 1	Recurrent Neural Networks (RNN)
	<ul> <li>RNN Explained   Characteristics of RNN</li> <li>Backpropagation</li> <li>Gradient problem (Vanishing vs. Exploding)</li> <li>Long Short-Term Memory Networks</li> </ul>
Session 2	<ul> <li>RNN vs LSTM with Google Stock Price</li> <li>Transform Data for Time Series</li> <li>MLP &amp; CNN for Time Series Forecasting</li> <li>LSTM &amp; CNN-LSTM for Time Series Forecasting</li> <li>Encoder-Decoder LSTM Multi-step Forecasting</li> </ul>
Session 3	<ul> <li>Sentiment Analysis with LSTM</li> <li>Motivations</li> <li>From logistic regression to neural networks</li> <li>Word representations</li> <li>Unsupervised word vector learning</li> <li>Backpropagation Training</li> <li>Learning word-level classifiers: POS &amp; NER</li> </ul>
Session 4 Session 5	<ul> <li>GANS: Generative Adversarial Networks</li> <li>Challenge of GAN Loss</li> <li>Standard GAN Loss Functions</li> <li>Alternate GAN Loss Functions</li> <li>Effect of Different GAN Loss Functions</li> <li>Python: Torch Library 2</li> <li>How to Develop PyTorch Deep Learning Models         <ul> <li>How to Develop an MLP for Binary Classification   Multiclass Classification</li> <li>How to Develop an MLP for Regression</li> </ul> </li> </ul>
Week 22	How to Develop a CNN for Image Classification  General Deployment Considerations & MLOps

Learning Objectives: In this Week, you will learn every aspect of how to put your models in production. You will learn all the steps and infrastructure required to deploy machine learning models professionally, and you will have at your fingertips, the sequence of steps that you need to follow to deploy a machine learning model, plus a project template with full code, that you can adapt to deploy your own models.

# Session 1 Avoiding False Discoveries | MLOps Introduction Preliminaries: Statistical Testing Modeling Null & Alternative Distributions What is MLOps? Machine learning industrialization challenges MLOps Motivation: High-level view MLOps challenges MLOps Components Machine Learning Life Cycle



7.7	<ul> <li>How does it relate to DevOps, AIOps, ModelOps, and GitOps?</li> </ul>
	Major Phases - what it takes to master MLOps.
	CI/CD in Production Case Study
Session 2	Introduction to ML and MLOps stages
	MLOps Maturity Model Detailed
	MLOps and stages
	<ul> <li>Versioning</li> </ul>
	o Testing Automation (CI/CD)
	o Reproducibility
	o Deployment Monitoring
Session 3	MLOps Architectures
	MLOps Architectures
	o Architectures - Open Source tools - Kubeflow, Apache Airflow, MLFlow, Metaflow,
	Kedro, ZenML, MLRun, CML
	Architectures - Cloud Native tools - AWS, GCP and Azure  List of tools involved in each stage (ML One tool accounts)
	<ul> <li>List of tools involved in each stage (MLOps tool ecosystem)</li> <li>Different Roles involved in MLOps (ML Engineering + Operations)</li> </ul>
Session 4	Introduction to CI/CD
	Introduction to CI and CD
	CI/CD challenges in Machine Learning
	Steps involved in the CI/CD implementation in ML lifecycle and workflow.
	A glimpse of popular Tools used in the DevOps ecosystem on the Cloud.
Session 5	Python: TensorFlow & Keras Library 1
	Overview of TensorFlow and TensorFlow libraries
	Use cases for a machine learning service.
	Using and applying your model
	Training your model   Testing your model
	Using TensorBoard to visualize model performance
W. 1 00/06	

#### Week 23/26

#### **Individual Capstone Project (1 Month)**

**Learning Objectives:** During the final week, you will be transitioned into full-time focus on your final, passionindividual capstone assignment that will require you to apply the tools you have learned in the program.

**Duration:** The capstone will run for ONE month starting from the 6th of January 2024 till the6th of February 2024.

**Grading:** The capstone project is a program requirement. To successfully complete the capstone project and receive the certificate, students must achieve "Meets Expectations" OR"Exceeds Expectations".



#### Week 27 | 28

#### **Data Science & AI Interview Preparation**

**Learning Objectives:** In this module, you will be introduced to 120+ Data Science Interview questions and answers to help you master the Data Science language and express yourself fluently. Topics covered:

- The Big Picture
- Optimization
- Data Pre-processing
- Sampling & Splitting
- Supervised Learning
- Unsupervised Learning
- Model Evaluation
- Ensemble Learning
- Business Applications

#### Week 29|30

#### **Data Science & AI 24 Hours Real Challenges**

**Learning Objectives:** In this module, you will be introduced to some 24-hour real take-home challenges used as hiring. process with some big companies:

- Capgemini Challenge
- SDSC Challenge
- Foot Locker Challenge
- Fatality Crashes Challenge
- PayPal Challenge

#### Week 31

#### **Data Science Soft Skill Preparation**

<u>Learning Objectives</u>: In this module, you will assess your Data Science soft skills in core topics of statistics, fundamentals of data science and programming, Machine Learning, and Neural Networks. The test is designed to help you identify your strengths and weaknesses.

#### Week 32|40

#### **Introduction to ML and MLOps stages**

<u>Learning Objectives</u>: InnovatiCS has developed a new course "CAP Preparation Course" as a bonus & great opportunity for all students at all experience levels to prepare for the Certified Analytics Professional (CAP®) certification or to refresh their knowledge and skills. In this module, you will explore all seven domains of the analytics practice as:



- Domain I: Business Problem Framing
- Domain II: Analytics Problem Framing
- Domain III: Working with Data
- Domain IV: Methodology Selection
- Domain V: Model Building
- Domain VI: Model Deployment
- Domain VII: Life-Cycle Management
- Mock exam



## **Certificate of Completion**

Upon successful completion of the program, Big Bang Data Science Institute grants a **verified/certified digital** certificate of completion to participants. This program is graded as pass or fail; participants must receive 80% to pass and obtain the certificate of completion.



This course has been certified by the CPD Certification Service as conforming to continuing professional development principles. After successful completion of the program, your verified digital certificate will be email to you in the name you used when registering for the program. All certificate images are for illustrative purpose only and may be subject to change at the direction of InnovatiCS.

The CPD Certification Service: The CPD Certification Service was established in 1996 and is the leading independent CPD accreditation institution operating across industry sectors to complement the CPD policies of professional and academic bodies.



## **Testimonials**

Here is another reason why you should trust **INNOVATICS** with your training. Watch this recording capturing **INNOVATICS** prestigiousAward from Internet 2.0 Conference - Dubai 2022, Internet 2.0 Conference **Dubai 2023**, Internet 2.0 Conference **Las Vegas** 2023.









## **Our Clients!**



Our diverse clients and projects have continued to provide our team with the opportunity to grow a unique skillset. The following are few of our recently completed programs for corporates or universities.

#### **OUR MOST RECENT CLIENTS INCLUDE:**































PROPOSAL FOR DATA SCIENCE GO



## **About INNOVATICS**

We are **INNOVATICS**, a holistic up-skilling platform driven by a unique, cohesive "Learn-Apply-Solve" framework. This innovative solution provides application-oriented immersive and interactive learning experience with extensive real-industry courses, cases, datasets and projects. It also ensures a blended pathway between industry and academia through simulation and context-ualisation.

**INNOVATICS** regularly presents at numerous conference workshops and until recently held regular monthly Meetups with industry experts as speakers.

We currently are a few multi-week, multi session courses that are live (then recorded) programs that participants have thoroughly enjoyed since we support our participants with almost endless one-on-one or group live support sessions.



## **Connect with a Program Advisor**

Have questions about the program or how it fits in with your career goals?

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