Sheth NKTT College of Commerce and Sheth JTT College of Arts, Thane (Autonomous)

Credit Structure: Post Graduate Programme

As per NEP-2020

(w.e.f. 2025-26)

Semester-I M.Sc. (Data Science)

Category	Course	Credit
	Essential Technologies for Data Science	4
Mandatory	Essential Technologies for Data Science Practical	2
	Data Analysis and Visualization	4
	Data Analysis and Visualization Practical	2
	Statistical Methods for Data Science	2
	SPARK Technologies (2 TH+2PR)	4
OE (Any One)	Retail Marketing Analytics	
- ()	(2 TH + 2 PR)	
	Sports Data Analytics (2 TH +2PR)	
	Research Methodology	4
	Total	22

Programme Name: M. Sc. (Data Science)	Semester: I
Course Category/Vertical: Mandatory	
Name of the Dept: Science and Technology	
Course Title: Essential technologies for Data science	
Course Code:	Course Level:6.0
Type: Theory	
Course Credit: 4 credits	
Hours Allotted: 60 Hours	
Marks Allotted: 100 Marks	
Course Objectives(CO):	
CO1: Learn Python basics including programming elements and functions for data handling.	s, data structures, control flow,
CO2: Understand core statistics concepts like probability, d and apply them using NumPy.	istributions, hypothesis testing,
CO3: Perform data wrangling and analysis using Pandas for summarizing datasets.	r cleaning, transforming, and
CO4: Gain skills in data visualization using Matplotlib and	R (ggplot2) and understand

Course Outcomes (OC):

different types of data and analysis methods.

Students should be able to understand the core concepts of programming before starting to write new programs. Students should be able to develop logic for Problem Solving.

To manage data by relying on data structures such as strings, arrays, files, lists, and dictionaries. Exemplify the various levels of decision making on a program and implement a mix of loops, functions, and control flow to extract information from a data structure.

Student should able to learn different programming techniques and tools related to data science.

Syllabus: NEP 2020 w.e.f 2024-25

Essential technologies for Data science

Unit No.	Content	Hours
	Module I	
Ι	 Unit 1: Introduction to Data science and Python a) Introduction to Data Science, data science life cycle, Applications, and advantages of Python over other programming languages b) What is Python? Why Should I learn Python? Installing Python How to execute Python program Writing your first program. c) Basic programming elements of Python-variables and constants, identifiers, Typecasting or Type Conversion in Python, indentation, comments, rules of writing identifiers, primitive data types, writing command line programs in python d) Operators in Python: Arithmetic operators, relational operators, Logical operators, Membership operators, Taking user input. Unit 2: Data structures and control flow a) Collection data structures in Python- List, tuples, dictionary, sets and strings b) Control flow- Sequential, Branching or Conditional, 	15
	 c) Condition new Sequencial, Drahening of Conditional, Iteration or Repetition, Modular or Subroutines Conditional and iteration statements: if else statements, loops, for loop and while loops c) User defined functions in Python- No Value Pass and No Return, Value Pass and No Return, Value Pass and Return, Function With default arguments, Function with variable arguments, Higher order functions, list comprehension 	
II	Unit 3:Statistics for Data Analysts	15
	 a) Permutations and combinations, probability, Descriptive statistics (mean, median, mode), point estimation, quartiles and boxplot, methods of dispersion, random variables and probability distribution b) Measures of shape- skewness, kurtosis, outlier detection, transformation (log, square root) 	

 d) Unit 4 a) b) c) 	 Ratio), continuous and discrete numeric data. Types of data analysis (descriptive, diagnostic, predictive and prescriptive analysis) Data wrangling using Pandas - Creating Series, Creating Data frame from dictionary, attributes, and method description of a data frame. Drop columns, add columns, add rows, iloc , loc, indexing and slicing data frames, selection with condition, group by summary operation, sorting operations Introduction to R IDE- components of R IDE, Basic data types in R, Data structures in R, data coercion, importing files, visualisation using ggplot2. 	
	sualisation using matplotlib- Components of a chart, line chart, chart, pie chart, sub plots.	
Total H	lours	60

References :-

Data Analysis with Pandas and Python by Boris Paskhaver, Manning Publications. Available at: <u>https://www.perlego.com/book/2881120/pandas-in-action-pd</u> Practical Statistics for Data Scientists: 50 Essential Concepts by Peter Bruce, Andrew Bruce, Peter Gedeck, O'Reilly Media, 2017 ISBN-10: 1491952962 ISBN-13: 978- 1491952962 Foundations of Statistics for Data Scientists With R and Python By Alan Agresti, Maria Kateri, CRC Press Taylor and Francis group, 2022

Programme Name: M. Sc. (Data Science)

Semester: I

Thane (W)

Course Category/Vertical: Mandatory

Name of the Dept: Science and Technology

Course Title: Essential technologies for Data Science Practical

Course Code:

Course Level:6.0

Type: Practical

Course Credit: 2 credits

Hours Allotted: 60 Hours

Hours Allotted. 60 Hours

Marks Allotted: 50 Marks

Course Objectives(CO):

- CO1 : To develop hands-on skills in Python programming by practicing arithmetic, relational, logical operations, loops, and working with core data structures.
- CO2: To gain experience in data analysis by importing datasets, performing univariate, bivariate, and multivariate analysis, and visualizing data using Python, R, or Excel.
- CO3: To apply statistical techniques such as hypothesis testing, ANOVA, and correlation analysis with appropriate visualizations like heatmaps.
- CO4 : To understand data wrangling and preprocessing through operations like data selection, filtering, updating, grouping, and sorting using tools like Pandas or spreadsheet software.

Course Outcomes (OC):

- Should be able to write basic programming in Python.
- Should be able to use Python data structures and able to use conditional and iterative control flow.
- Should be able to demonstrate descriptive, diagnostic, and inferential statistics using, Python, R or Excel (use Data analysis tool pack in Excel or Data analyzer tool in Microsoft office 365)
- Perform basic data wrangling using R or Pandas
- Perform data visualization using R or Pandas

Syllabus: NEP 2020 w.e.f 2024-25

Unit No.	Name of Practical	Hours
1	Write a Python program to accept inputs from users and	

	perform arithmetic operations.	
2	Write a program to demonstrate relational and logical	
	operators in Python.	
3	Write a Python program to demonstrate usage of loops. Use both	
	for and while loops to distinguish between them. [e.g., Reversing	
	the digits of a number without converting to String]	
4	Demonstrate the use of data structures list, sets, dictionary.	
5	Import a dataset and perform univariate analysis on the numeric	
	columns	
6	Demonstrate Hypothesis testing, and ANOVA using a dataset [use	
	Python, R or Excel]	
7	Demonstrate correlation analysis. Use heatmap for	
	visualization.	
	Write inferences.	
8	Import a csv or Excel dataset and demonstrate data wrangling, view shape, dimension, column names of the dataset, ways to select data using column number, column names, simple and compound conditional selection, update and modify dataset.	
9	Demonstrate group by summary operations and sorting techniques.	
10	Perform univariate, bivariate and multivariate analysis using visualization techniques in Python, R or Excel	
		60

of Arts, (Autonomous) Thane (W)

Programme Name: M. Sc. (Data Scidence)	Semester: I
Course Category/Vertical: Mandatory	
Name of the Dept: Science and Technology	
Course Title: Data Analysis and Visualization	
Course Code:	Course Level:6.0
Type: Theory	
Course Credit: 4 credits	
Hours Allotted: 60 Hours	
Marks Allotted: 100 Marks	

Course Objectives(CO):

- CO1 : Understand the basics of data analysis, including the data science process, types of data, sources, file formats, and tools used by data professionals.
- CO2: Gain practical skills in data wrangling, such as importing data, cleaning, transforming, merging datasets, and handling missing or unstructured data.
- CO3: Learn to visualize data effectively using tools like Tableau by creating various charts, dashboards, and adding interactivity for deeper insights.
- CO4: Develop storytelling skills with data using Power BI by designing interactive reports, customizing visuals, and presenting data-driven insights clearly.

Course Outcomes (OC):

- Understand various data formats, sources and storage mechanisms.
- Handle missing data and manage data wrangling and manipulation
- Create data visualization and report making using various software tools
- Demonstrate the visualizations and make interpretations
- Create a data story using various software tools.

Unit No.	Content	Hours
Ι	 Unit 1: Introduction to Data Analysis Data Analysis - Exploratory Data Analysis and Data Science Process Responsibilities of a Data Analyst - Data Analytics vs. Data Analysis - Types of Data - Understanding Different Types of File Formats - Sources of Data - Languages for Data Professionals - Overview of Data Repositories - Data Marts, Data Lakes, ETL, and Data Pipelines - Foundations of Big Data - Identifying Data for Analysis Unit2: Data Wrangling Data Sources - How to gather and Import Data - Data Loading, 	15

Storage and File Formats - Reading and Writing Data in Text Format, Web Scraping, Binary Data Formats, interacting with Web APIs, Interacting with Databases – Data Wrangling - Hierarchical Indexing, Combining and Merging Data Sets Reshaping and Pivoting - Tools for Data Wrangling - Data Cleaning and Preparation - Handling Missing Data, Data Transformation, String Manipulation	
	15
Unit 3: Data Visualization Intro to data visualization - Introduction to Visualization and Dashboarding Software - Visualization Tools - Getting started with Tableau Desktop – Connecting to the dataset - Creating charts – Creating common visualizations (bar charts, line charts etc.) – Filtering and sorting data - Adding Titles, Labels, and descriptions - Publish your work to Tableau Cloud - Interactivity with text and visual tooltips - Interactivity with actions (filter, highlight, URL) – Assembling dashboards from multiple charts	
Unit 4: Story Telling	
Introduction to Power BI - Understanding Desktop - Understanding Power BI Report Designer - Report Canvas, Report Pages: Creation, Renames - Report Visuals, Fields and UI Options - Experimenting Visual Interactions, Advantages - Reports with Multiple Pages and Advantages - Pages with Multiple Visualizations - PUBLISH Options and Report Verification in Cloud - Adding Report Titles. Report Format Options - Introduction to data storytelling - Creating a data story	
Total Hours	60
Svllabus: NFP 2020 w e f 2024-25	L

Syllabus: NEP 2020 w.e.f 2024-25

References :-

- 1. Python for Data Analysis: Data Wrangling with Pandas, NumPy and IPython by McKinney, W., 2nd edition. O"Reilly Media, 2017
- 2. Doing Data Science: Straight Talk from the Frontline by O"Neil, C., & Schutt, R, O"Reilly Media, 2013
- 3. The Big Book of Dashboards by Steve Wexler, Jeffrey Shaffer, Andy Cotgreave, John Wiley & Sons, 2017
- 4. Practical Tableau by Ryan Sleeper, O"Reilly Media, 2018
- 5. Power BI. Book-1, Business Intelligence Clinic: Create and Learn by Roger F Silva, 2018

6. Introducing Microsoft Power BI by Alberto Ferrari and Marco Russo, Microsoft Press, Washington, 2016

	e Name: M. Sc. (Data Scidence)	Semester: I
Course Cat	tegory/Vertical: Mandatory	
Name of th	ne Dept: Science and Technology	
Course Tit	le: Data Analysis and Visualization P	ractical
Course Co	de:	Course Level:6.0
Type: Prac	tical	
Course Cre	edit: 2 credits	
Hours Allo	tted: 60 Hours	
Marks Allo	otted: 50 Marks	
Course Ob	jectives(CO):	
fron CO2 : To a tran CO3: To de buile CO4: To er visua	n various formats, web scraping, and in pply data wrangling and cleaning meth sforming data, and performing string n evelop skills in data visualization by cr ding interactive dashboards using tools reate and present data-driven stories the alizations, and publish them using Tabl	nods such as handling missing values, nanipulation for analysis. eating charts, applying filters and sorts, and like Tableau and Power BI. rough well-structured reports and
	utcomes (OC):	1. <u>1</u> . <u>1</u> .
•	Handle missing data and manage data	wrangling and manipulation
	с с	
•	Create data visualization and report n	naking using various software tools
•	с с	aking using various software tools ake interpretations

Syllabus: NEP 2020 w.e.f 2024-25

Unit No.	Content	Hours
1	Implement Data Loading, Storage and File Formats. Read data and store them in text format.	
2	Implement the code to interact with Web APIs and to perform web scrapping.	
3	Demonstrate Data Cleaning and Preparation.	
4	Implement Data wrangling on a data set.	
5	Demonstrate the handling of missing data and string manipulation.	
6	Create common charts with title, labels and descriptions using Tableau.	
7	Perform sorting and filtering using tableau, create visualizations and publish it on Tableau Cloud.	
8	Perform data visualization using Power BI.	
9	Create reports using Power BI.	
10	Create a data story in Tableau or power Bl.	
		60

Progra	ramme Name: M. Sc. (Data Science)	Semester: I
	se Category/Vertical: Mandatory	
Name	e of the Dept: Science and Technology	
Course	se Title: Statistical Methods for Data Science	
Course	se Code: C	ourse Level:6.0
Type: '	Theory	
	se Credit: 2	
Hours	s Allotted: 30 Hours	
Marks	s Allotted: 50 Marks	
2. 3.	 To understand foundational concepts in statistics and infer testing, error types, confidence intervals, and power analy To apply various statistical tests such as t-tests, ANOVA, or nonparametric methods to analyze and interpret data. To learn multivariate statistical techniques, including linear MANOVA, discriminant analysis, and their applications in To explore advanced techniques like Principal Component and Cluster Analysis 	rsis. correlation analysis, and ar and logistic regression, n data science.
Course	 se Outcomes (OC): Students will gain a solid understanding of foundation including probability, sampling distributions, hy confidence intervals. They will learn the principles statistical analysis. Students will learn how to apply statistical methods context of data science. They will become profici techniques such as regression analysis, analysis of va- square tests, and non-parametric tests. Students will gain proficiency in using statistical soft 	ypothesis testing, and and techniques used in s to analyze data in the ient in using statistical ariance (ANOVA), chi-

- and automate data analysis processes.
- Students will enhance their critical thinking skills and ability to solve problems using statistical methods.

I MODULE - I: 15 Unit 1: Introduction to Applied Statistics 15 The Nature of Statistics and Inference, What is "Big Data"?, Statistical Modelling, Statistical Significance Testing and Error Rates, Simple Example of InferenceUsing a Coin, Statistics is for Messy Situations, Type I versus Type II Errors, Point Estimates andConfidence Intervals, Variable Types, Sample Size, Statistical Power, and Statistical Significance, TheVerdict on Significance Testing, Training versus Test Data. Means, Correlations, Counts: Drawing Inferences: Computing z and Related Scores, Statistical Tests, Plotting Normal Distributions, Correlation Coefficients, Evaluating Pearson"s r for StatisticalSignificance, Spearman"s Rho: A Nonparametric Alternative to Pearson. Tests of Mean Differences: t-Tests for One Sample, Two Sample t-Test, Paired-Samples t- Test Categorical Data: Binomial Test, Categorical Data Having More Than Two Possibilities. Power Analysis and Sample Size Estimation: Power for t-Tests, Power for One-Way ANOVA, Power for Correlations. Analysis of Variance: Fixed Effects, Random Effects, Mixed Models, Introducing the Analysis of Variance (ANOVA), Performing the ANOVA, Random Effects ANOVA and Mixed Models, One-Way Random Effects ANOVA II Unit 2: Multivariate Techniques 15 Simple and Multiple Linear Regression, Hierarchical Regression and the Generalized Linear Model, Predicting Probabilities, Multiple Logistic Regression, Training Error Rate Versus Test Error Rate. Multivariate Analysis of Variance (MANOVA) Discriminant Analysis; Multivariate Tests of Significance, Example of MANOVA, Outliers, Homogeneity of Covariance Matrices, Linear Discriminant Function Analysis, Theory of Discrim	Unit No.	Content	Hours
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Discriminant Function Analysis, Theory of Discriminant Analysis, Predicting Group Membership, Visualizing		• •	
Analysis, Predicting Group Membership, Visualizing			
		• •	
Separation.		Separation.	
Principal Component Analysis: Principal Component		*	
Analysis Versus Factor Analysis, Properties of Principal			

Components, Component Scores, How Many Components to Keep? Exploratory Factor Analysis, Common Factor Analysis Model, Factor Analysis Versus Principal Component Analysis on the Same, Initial Eigenvalues in Factor Analysis, Rotation in Exploratory Factor Analysis, Estimation in Factor Analysis. Cluster Analysis: k-Means Cluster Analysis, Minimizing Criteria, Example of k-Means Clustering Hierarchical Cluster Analysis, Why Clustering Is Inherently Subjective. Nonparametric Tests: Mann– Whitney U Test, Kruskal– Wallis Test, Nonparametric Test for Paired Comparisons and Repeated	
Total Hours	30

References:

- Gupta S. C., Kapoor V. K.: Fundamentals of Mathematical Statistics; Tenth Edition.Sultan Chand &Sons. (2000)
- 2. Johnson, R.A., Wichern, D.W.: Applied Multivariate Statistical Analysis, Prentice-Hall, NewJersey, 2002.
- 3. Draper, N. R. and Smith, H. (1998), Applied Regression Analysis (John Wiley), Third Edition.
- 4. Purohit, S. G. Gore, S.D. and Deshmukh, S.R. (2015). Statistics using R, second edition. Narosa Publishing House, New Delhi.
- 5. Daniel W. W.: Applied Non-Parametric Statistics, First edition Boston-Houghton Mifflin Company.

Programme Name: M.Sc (Information Technology)	Semester: I	
Course Category/Vertical: Elective		
Name of the Dept: Science and Technoloy		

Course Title: SPARK Technologies

Course Code:

Course Level:6.0

Type: Theory

Course Credit: 2 credits

Hours Allotted: 30 Hours Marks Allotted: 100 Marks

Course Objectives(CO):

Course Outcomes (OC):

- Understand the concept of SPARK Technologies and its implementation
- Understand the concept of RDD
- Understand the implementation of SPARK SQL, GraphX, Performance Tuning.
- Use Sparks Resilient Distributed Datasets to process and analyze large data sets across many CPUs
- Understand how the GraphX library helps with network analysis problems
- Understand how Spark SQL lets you work with structured data

Syllabus: NEP 2020 w.e.f 2024-25

Unit No.	Content	Hours
	Module I	
	MODULE - I: Basics of SPARK	
	Technologies Unit 1:	
	Introduction to SPARK	
	Technologies	
	Components of the Spark unified stack, Features of Spark, Spark Web UI, an introduction	
Ι	to RDDs - Resilient Distributed Datasets, Launching and using Spark"s Scala and Python shell, Spark Context, Spark Ecosystem, In-	

Memory data – Spark, Creating,Loading and Saving RDD, Transformations in RDD, Actions in RDD, Key-Value Pair RDD, Map Reduce and Pair RDD operations RDD Partitions	
Unit 2: Implementation of SPARK Technologies Spark Applications vs. Spark Shell, Creating Spark Context, Building a Spark Application, Spark and Hadoop Integration-HDFS, Handling Sequence File, Spark RDD- RDD Lineage, RDD Persistence Overview, Distributed Persistence. Spark Streaming, ML library for Spark, Working with Statistics, SPARK SQL, GraphX, Performance Tuning.	
Total Hours	

References:

- 1) Learning Spark: Lightning-Fast Data Analytics 2nd Edition, by Jules S. Damji,Brooke Wenig, Tathagata Das, Denny Lee, O"Reilly, 2020
- 2) Apache Spark Machine Learning Blueprints 1st Edition, Kindle Edition by Alex Liu, Packt Publishing, 2016
- 3) Apache Spark 2.x Cookbook: Cloud-ready recipes for analytics and data science 2nd Edition, by Rishi Yadav, Packt Publishing, 2017

Programme Name:	M.Sc (Data Science)	Semester: I	
Course Category/W	ertical: Elective		
Name of the Dept:	Science and Technology		
Course Title: SPA	RK Technologies Practical		
Course Code:		Course Level:6.0	
Type: Practical			
Course Credit: 2 cr	redits		
Hours Allotted: 60	Hours		
Marks Allotted: 50	Marks		
Course Objectives	s(CO):		
Course Outcomes	(OC):		
• Understand the concept of SPARK Technologies and its implementation			
• Unders	• Understand the concept of RDD		

• Understand the implementation of SPARK SQL, GraphX, Performance Tuning

Unit No.	Content	Hours
1	Installation of Apache Spark	
2	Spark Basics and RDD interface	
3	Filtering RDDs, and the Minimum Temperature by Location Example	
4	Counting Word Occurrences using flatmap()	
5	Executing SQL commands and SQL-style functions on aDataFrame	
6	Implement Total Spent by Customer with DataFrames	
7	Use Broadcast Variables to Display Movie Names Instead of ID Numbers	
8	Create Similar Movies from One Million Rating	
9	Using Spark ML to Produce Movie Recommendations	
10	Use Windows with Structured Streaming to Track Most-Viewed URLs (Spark Streaming)	

References:

Programme N	Name: M.Sc (Data Science) Semester: I	
Course Categ	gory/Vertical: Elective	
Name of the l	Dept: Science and Technology	
Course Title:	Retail Marketing Analytics	
Course Code:	: Course Level:6.0	
Type: Theory	y	
Course Credit	t: 2 credits	
Hours Allotte	ed: 30 Hours	
Marks Allotte	ed: 50 Marks	
Course Obje	ectives(CO):	
Course Outc	comes (OC):	
• 0	Inderstand the importance of marketing analytics for forward	
lo	ooking and systematic allocation of marketing resources	
• Know how to use marketing analytics to develop predictive		
marketing dashboard for organization		
• A	Analyze data and develop insights from it to address strategic marketing challenges	

Syllabus: NEP 2020 w.e.f 2025-26

Unit No.	Content	Hours
	Module I	
Ι	MODULE I Unit 1: Introduction to Marketing Analytics and Exploratory Data analytics using R a) Course Introduction • why marketing analytics? • course description and learning objectives b) Marketing Analytics Overview • how analytics can assist marketing decision-making • the framework of marketing optimization c) Tabulate and Summarize data • what cleaned data looks like • simple histogram plot • use histogram and boxplot to inform data distribution d) Visualize data • elements of data visualization • histogram, scatter plot, line plot, bar charts, line fits with the ggplot() function	
	 Unit 2:Marketing Campaigns - Experiment Design, Customer Lifetime Value (CLV) and Cohort Analysis a) Design and Conduct Experiments design experiments, examples randomization/sample splitting conduct experiments b) Assess Experiment Outcome Using Hypothesis Testing why hypothesis testing for experiment outcomes terminologies for hypothesis testing how does hypothesis testing work power calculation conduct hypothesis testing in R c) Calculate and Predict CLV calculate CLV typical frameworks in predicting CLV using linear regression and logistic regression to predict CLV d) CLV Analysis and Cohort Analysis Introduction to Experiment 	
	Total Hours	

References:

- 1. Hands-on Data Science for Marketing by Yoon Hyup Hwang, Packt Publishing, 2019
- 2. Retail Analytics: The Secret Weapon by Emmett Cox,1st edition, Weily, 2011
- 3. Cutting Edge Marketing Analytics: Real World Cases and Data Sets for Hands on Learning by Venkatesan Rajkumar, Farris Paul and Ronald Wilcox, Pearson FT Press, 2014
- 4. Marketing Analytics: A Practical Guide to Real Marketing Science by Grigsby Mike,Kogan Page, 2015

Programme Name: M.Sc (Data Science)	Semester: I
Course Category/Vertical: Elective	
Name of the Dept:	
Course Title: Retail Marketing Analytics P	ractical
Course Code: MSR107	Course Level:6.0
Type: Practical	
Course Credit: 2 credits	
Hours Allotted: 60 Hours	
Marks Allotted: 100 Marks	
Course Objectives(CO):	
Course Outcomes (OC):	
To Learn working and analyzing	g with marketing data
• To develop predictive marketing	g dashboard for organization
• Understand the concept of hy assessing experiment outcomes	pothesis testing and its role in

Unit No.	Content	Hours
1	Learn how to tabulate and summarize marketing data using R.	
	• Clean and preprocess the marketing data.	
	• Generate a simple histogram plot to visualize data distribution.	
	• Use tabulation and summary functions to gain insights from the data.	
	• Interpret the findings and discuss the implications for marketing analysis.	
2	Gain proficiency in visualizing marketing data using R.	
	• Understand the key elements of data visualization.	
	• Create various visualizations such as histograms, scatter	
	plots, line plots, and bar charts using the ggplot() function in R.	
	Apply appropriate visualization techniques to effectively communicate marketing insights.	
3	Design and conduct experiments for marketing campaigns.	
5	 Learn about experimental design and its application in 	
	marketing.	
	• Design experiments using examples from marketing	
	scenarios.	
	• Implement randomization and sample splitting techniques.	
	Conduct the experiments and collect relevant data for analysis.	
4	Understand the concept of hypothesis testing and its role in	
	assessing experiment outcomes.	
	• Explore the purpose of hypothesis testing in analyzing experiment results.	
	• Familiarize with key terminologies related to hypothesis testing.	
	 Learn the process of hypothesis testing and power 	
	calculation.	
	 Conduct hypothesis testing using R to evaluate experiment outcomes. 	
5	Calculate and predict Customer Lifetime Value (CLV).	
	• Calculate CLV using different approaches and frameworks.	
	• Explore predictive modeling techniques such as linear	
	regression and logistic regression for CLV prediction.	

	Assess the accuracy and reliability of CLV predictions.	
6	Apply CLV analysis and cohort analysis in marketing analytics.	
0	 Analyze CLV data and identify patterns and trends. 	
	• Perform cohort analysis to segment customers based on their	
	behavior or characteristics.	
	Interpret the results of CLV analysis and cohort analysis to derive	
	actionable insights for marketing strategies.	
7	Extract data from social media platforms and perform analysis	
	to gain insights into customer behavior and preferences.	
	• Utilize Python libraries like BeautifulSoup and requests to	
	scrape data from social media platforms.	
	Clean and preprocess the scraped data.	
	• Analyze the data to identify trends, sentiment analysis, or	
	customer engagement metrics.	
	Visualize the findings using appropriate charts or graphs.	
8	Analyze customer purchasing patterns and build a recommender	
_	system based on market basket analysis.	
	• Use transactional data to identify frequently occurring item	
	sets using association rule mining algorithms.	
	• Calculate support, confidence, and lift for the identified item	
	sets.	
	• Build a recommendation engine using collaborative filtering	
	techniques.	
	Evaluate the performance of the recommender system and make	
	recommendations based on customer preferences.	
9	Segment customers based on their recency, frequency, and	
	monetary value (RFM) to better target marketing	
	efforts.	
	• Analyze customer transaction data to calculate RFM scores.	
	• Segment customers into different groups using clustering	
	algorithms such as k-means or hierarchical clustering.	
	Perform descriptive analysis on each customer segment to understand their characteristics.	
	• Develop targeted marketing strategies for each segment based on their RFM profiles.	
10	Conduct A/B testing to evaluate the impact of different	
	marketing strategies and make data-driven decisions.	
	• Design and implement A/B tests for marketing campaigns	
	using randomized assignment.	
	Collect relevant data and perform statistical analysis to	
	compare the performance of different strategies.	
	Calculate key metrics such as conversion rates, click-through	

rates, or revenue. Interpret the results and provide recommendations for optimizing	
marketing campaigns based on the findings.	

Programme Nan	ne: M.Sc (Data Science)	Semester: I
Course Category	V/Vertical: Elective	
Name of the Dep	ot: Science and Technology	
Course Title: Sp	orts Data Analytics	
Course Code: M	SR107	Course Level:6.0
Type: Theory		
Course Credit: 2	credits	
Hours Allotted:	30 Hours	
Marks Allotted:	50 Marks	
Course Objectives(CO):		
Course Outcomes (OC):		
• Und	erstand the role and importa	nce of data analytics in sports
• Dev	elop skills in collecting, clea	ning, and managing sports data
• Gair	n proficiency in using statisti	cal analysis techniques to analyze sports data
• App	ly data visualization method	s to present sports data effectively
• Lea	n how to apply predictive m	odeling techniques to sports data
• Exp	lore the use of machine learn	ing algorithms in sports analytics

• Understand ethical considerations and challenges in sports data analytics

Unit No.	Content	Hours
	Module I	
	MODULE I Unit 1: Fundamentals of Sports Data Analytics	
	 A. Introduction to Sports Data Analytics Overview of sports data analytics Evolution of analytics in sports Importance and applications of sports data analytics 	
Ι	 B. Data Collection and Preprocessing Sources of sports data Data collection methods Data cleaning and preprocessing techniques Database management for sports analytics C. Data Visualization for Sports Analytics Principles of data visualization 	
	 Visualization tools and libraries Creating effective visualizations for sports data Interactive dashboards for sports analytics D. Statistical Analysis in Sports Descriptive statistics for sports data Hypothesis testing in sports analytics Regression analysis in sports Analysis of variance (ANOVA) in sports 	
	 D. Predictive Modeling in Sports Analytics Introduction to predictive modeling Feature selection and engineering for sports data Linear regression models for sports prediction Classification models for sports outcomes E. Machine Learning in Sports Analytics Overview of machine learning algorithms Decision trees and random forests in sports analytics Support vector machines for sports prediction Neural networks and deep learning in sports 	

Total
 F. Advanced Topics in Sports Data Analytics Sports performance analysis Player tracking and motion analytics Sports marketing and fan engagement analytics Sports injury prediction and prevention G. Sports Business Analytics Revenue generation and marketing in sports Fan engagement and customer analytics
analytics

Reference Books:

- Sports Analytics: A Guide for Coaches, Managers, and Other Decision Makers by Benjamin C. Alamar, Columbia university press, 2013
- 2. Sports Analytics and Data Science: Winning the Game with Methods and Models by Thomas Miller, 1st edition, Pearson FT Press, 2015
- 3. Sports Analytics: Analysis, Visualisation and Decision Making in Sports Performanceby Daniel Memmert, Tim McGarry, and Tony Reilly, 2018
- 4. Cricket Analytics: Analytics and Data Science in Cricket by Tapan Bagchi and S. Raghunathan
- 5. Machine Learning using Python by Manaranjan Pradhan and U. Dinesh Kumar, Weily , 2020

URL for Online Study Material -

- 1. www.coursera.org
- 2. www.kaggle.com
- 3. www.datacamp.com
- 4. www.sportsanalyticsinstitute.com

Programme Name: M.Sc (Data Science)	Semester: I	
Course Category/Vertical: Elective		
Name of the Dept: Science and Technology		
Course Title: Sports Data Analytics Practical		
Course Code:	Course Level:6.0	
Type: Practical		

Course Credit: 2 credits

Hours Allotted: 60 Hours

Marks Allotted: 50 Marks

Course Objectives(CO):

Course Outcomes (OC):

- Understanding of Sports Data Analytics
- Proficiency in Data Manipulation and Exploratory Data Analysis
- Player Performance/ Team Performance Analysis
- Predictive Modeling
- Data Visualization and Reporting

Unit No.	Content	Hours
1	Exploratory Data Analysis	
	 Perform exploratory data analysis on a cricket dataset, analyzing variables such as number of matches, runs, not outs, wickets, etc. Visualize the distribution of player performance metrics using histograms, box plots, or scatter plots. Investigate the relationship between player age and performance 	
2	metrics using correlation analysis. Batting Performance Analysis	
	 Analyze batting performance in a Cricket dataset, calculating metrics such as batting average, strike rate, and runs scored by players. Identify top-performing batsmen based on performance metrics and compare their performance against different opponents or in specific conditions. 	
3	 Bowling Performance Analysis Analyze bowling performance in the Cricket dataset, 	

	 calculating metrics such as bowling average, economy rate, and wickets taken by players. Identify top-performing bowlers based on performance metrics and analyze their performance against different teams or in various match
	situations
4	Performance Comparison • Compare the scoring averages of top-performing batsman in different seasons. Analyze the runs scoring (strike rate of Batting) of players from various teams in a specific league
5	Player Position Analysis Calculate position-specific performance metrics and compare players
	within each position.
6	 Injury Analysis Investigate the relationship between player injuries and their subsequentperformance using historical injury and performance data. Identify patterns and trends in the data to determine the impact of
	injuries on playerperformance and team success.
7	Team Analysis • • Analyze the impact of toss on a team's overall scoring and winning percentage. • Study the relationship between batting averages of players and their •
	team's win-loss record
8	Sports Revenue Analysis • • Analyze revenue generation in sports organizations by examining factors such as ticket sales, merchandise sales, and sponsorship deals. • Identify key drivers of revenue and provide recommendations for •
	maximizing financial performance.
9	Predictive Modeling • Build a regression model to predict the number of runs scored by players based ontheir historical performance data. Develop a classification model to predict the outcome of match
	based on team"sstatistics.
10	Visualization and Reporting: (Mini-Project)
	Prepare a comprehensive report summarizing the findings of the analysis and providing actionable insights for sports teams or

	organizations.	
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Programme Name: M.Sc (Data Science) Semester: I

Course Category/Vertical:

Name of the Dept: Science and Technology

Course Title: Research Methodology

Course Code:

Course Level:6.0

Type: Theory

Course Credit: 4 credits

Hours Allotted: 60 Hours

Marks Allotted: 100 Marks

Course Objectives(CO):

Course Outcomes (OC):

- Formulate clear and relevant research questions and objectives in the context of Data Sciences.
- Choose appropriate research designs, sampling methods, and data

collection techniques for data-driven research projects.

- Gather, preprocess, and manage datasets, ensuring data quality and ethical handling.
- Apply statistical analysis, machine learning, and deep learning techniques to analyze and extract insights from data in Data Sciences.
- Effectively communicate research findings through research papers and presentations tailored for the Data Sciences community.
- Understand and adhere to ethical principles and data privacy regulations when conducting research involving sensitive data.
- Prepare and submit research manuscripts to conferences and journals, understanding the peer-review process in Data Sciences.
- Analyze and extract insights from large-scale datasets using big data analytics tools and techniques.
- Recognize and address ethical and security concerns related to data privacy and data protection in research.

Unit No.	Content	Hours

	Module I	
Ι	 MODULE – I Unit 1: Introduction to Research Methodology for Data Sciences Understanding Research in Data Sciences: Definition and significance of research in Data Sciences. Definition and practical research, applied research, and practical research, Research approaches: Quantitative vs. Qualitative, The role of research in Data Sciences. Research Process and Ethics: Steps in the research process in Data Sciences, Research ethics and responsible conduct, Formulating research questions and objectives, Identifying research problems in Data Sciences, Ethical considerations in research, including data privacy and security. Research Design and Sampling for Data Sciences: Principles of research design in Data Sciences, Types of research design, Sampling techniques in Data Sciences, Choosing the appropriate research design for data-driven research, Ethical considerations in sampling and data collection, 	
	 including informed consent. MODULE - II Unit 3: Data Analysis and Modeling for Data Sciences Statistical Analysis for Data Sciences: Inferential statistics for hypothesis testing, Regression analysis and correlation, Machine learning concepts for predictive modeling. Machine Learning and Deep Learning: Introduction to machine learning and deep learning algorithms, Model selection and evaluation, Big Data Analytics: Understanding big data and its challenges, Tools and frameworks for big data processing (e.g., Hadoop, Spark), Analyzing and extracting insights from large datasets. 	
	 Unit 4: Research Communication, Publishing, and Data Privacy Research Paper Writing and Presentation: Structure of a research paper, Writing the abstract, introduction, literature review, methodology, and results sections, Effective research paper presentations. Publishing Research in Data Sciences: Choosing the right conferences and journals, The peer-review process in Data Sciences, Preparing manuscripts for submission, Ethical considerations in publishing, including data privacy and security in publications. Ethical Considerations in Data Privacy and Security: Data privacy 	
	regulations and compliance, Ethical considerations in data	
	anonymization and de- identification, Securing research data and protecting sensitive information.	

References :

1. Research Methodology: Methods and Techniques, C.R. Kothari, New Age International