

# **Syllabus**

# **Honours Programme in**

# **Data Science & Analytics**

(Offered by Department of Computer Engineering)

# **From**

# Academic Year 2021-22

# **Revision 2**

(Approved in Academic Council meeting dated )



# K J Somaiya College of Engineering, Mumbai-77

( A Constituent College of Somaiya Vidyavihar University)

K. J. Somaiya College of Engineering, Mumbai -77 (A Constituent College of Somaiya Vidyavihar University)

### **Honours' Degree Programme in Data Science & Analytics**

Offered by Department of Computer Engineering

### **Introduction:**

In today's data-driven society, Data Science provides a foundation for problem solving that impacts virtually all areas of the economy, including science, engineering, medicine, banking, finance, sports and the arts. Data science is an interdisciplinary field that focuses on analysing large amounts of data to identify inherent patterns, extract underlying models, and make relevant predictions.

Data processing and analytics converts raw data into format which can be analysed and interpreted for a variety of purposes. It focuses on data processing techniques and algorithms for representation of data in a meaningful way for human intelligence. The focus is also on innovative and intelligent ways of handling data which may be unstructured, high in volume, of different variety and analyzing data for various purposes.

The data science and analytics programme is designed to prepare students in wide disciplines who want to gain practical know-how of data analytics methods as it relates to their field of interest. It is designed to empower them to employ computational thinking and data science tools to solve practical business problems. The coursework consists of courses that cover the spectrum of Data Science to equip the students with knowledge of data analysis techniques and data-centric computation to address problems that require large data.

### **Objectives:**

- Applications of principles of Data Science to the analysis of diverse problems.
- Use software tools and algorithms from statistics, applied mathematics, Computer Science to model and analyze real-world data, communicate findings, and effectively present results using data visualization techniques.
- Deployment of latest tools and technologies to analyze large amounts Data.
- Understand the ethical practices that are importantly and inevitably tied to data-driven decision-making.

### **Learning Outcomes of the Honours' Degree Programme:**

At the successful completion of this programme an engineering graduates will be able to

- Apply principles of Data Science for analytics to diverse problems.
- Demonstrate the use of various tools in the domain of Data Science for data visualization.
- Discuss ethical practices related to data-driven decision-making.

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• Implement solutions to data analysis problems using latest tools and technologies.

**Assessment Methods:** Evaluation is done by a variety of tools including Open book tests, MCQs (multiple choice questions), Study of research papers, Internal Assessment tools and End Semester Examinations etc. Mini-Projects are offered in courses also to encourage project based learning among students.

Acronyms used in syllabus document						
Acronym	Definition					
CA	Continuous Assessment					
ESE	End Semester Exam					
IA	Internal Assessment					
O	Oral					
P	Practical					
P&O	Practical and Oral					
TH	Theory					
TUT	Tutorial					
TW	Term work					
ISE	In-semester Examination					
CO	Course Outcome					

### Acronyms used in Course code e.g. 116h54C301

<b>Position of Digit</b>	Acronym	Definition						
1	1	First revision SUV KJSCE 2020						
2	16	KJSCE						
3	h	Honour Degree Program						
4	55	Data Science & Analytics						
5	C	Core Course						
	L	Laboratory Course						
	T	Tutorial						
	P	Project Based Course						
6	1/2/3/4	Semester Number						
7	01/02/03	Course Number						

### **Credit Scheme**

Course Code	Course Name	Teaching Scheme (Hrs.) TH – P – TUT	Total (Hrs.)	Credits Assigned TH – P – TUT	Total Credits	Suggested semester of Honours' degree
116h54C301	Data Visualization	3-0-0	03	3-0-0	03	III
116h54L301	Data Visualization	0 - 2 - 0	02	0 - 1 - 0	01	III
116h54C401	Applied Data Science	3-0-0	03	3-0-0	03	IV
116h54L401	Applied Data Science	0 - 2 - 0	02	0 - 1 - 0	01	IV
116h54C501	Data Analytics	3-0-0	03	3-0-0	03	V
116h54L501	Data Analytics	0 - 2 - 0	02	0 - 1 - 0	01	V
116h54C601	Advanced Data Mining	3-0-0	03	3-0-0	03	VI
116h54C701	Advanced Machine Learning	3-0-0	03	3-0-0	03	VII
116h54P801	Applied Project/ Internship	0 - 4 - 0	04	0-2-0	02	VIII
	Total	15 – 10 – 0	25	15 – 5 – 0	20	

### **Examination Scheme**

Course	Course Name	Examination Scheme							
Code		Marks							
		C	A	ESE	TW	O*	P	P&	Total
		ISE	IA					0	
116h54C301	Data Visualization	30	20	50	-	-	-	-	100
116h54L301	Data Visualization	-	-	-	25	25	-	-	50
116h54C401	Applied Data Science	30	20	50	-	-	-	-	100
116h54L401	Applied Data Science	-	-	-	25	25	-	-	50
116h54C501	Data Analytics	30	20	50	-	-	-	-	100
116h54L501	Data Analytics	-	-	-	25	25	-	-	50
116h54C601	Advanced Data Mining	30	20	50	-	-	-	-	100
116h54C701	Advanced Machine Learning	30	20	50	-	-	-	-	100
116h54P801	Applied Project/ Internship	-	-	-	50	50	-	-	100
	Total	150	100	250	125	125			750

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<b>Course Code</b>		Course Title						
116h54C301		Data Visualization						
	,	TH P TUT Total						
Teaching Scheme(Hrs.)			-	-			03	
Credits Assigned		03						03
				Marks				
Examination	<b>Examination</b> CA		ESE	(DXX)		P	P&O	Total
Scheme	ISE	IA	LSE	TW	O	r	P&O	1 Otai
	30	20	50					100

### Course prerequisites (if any):

Basics of statistics, database and data analysis

### **Course Objectives**

- Employ best practices in data visualization to develop charts, maps, tables, and other visual representations of data
- Use visualization tools to conduct data analysis, especially exploration of an unfamiliar dataset.
- Create compelling, interactive dashboards to combine several visualizations into a cohesive and functional whole.
- Use data visualizations, dashboards and Stories to support relevant communication for diverse audiences.

### **Course Outcomes**

### At the end of successful completion of the course the student will be able to

CO1	Learn how to locate and download datasets, extract insights from that data and present their findings in a variety of different formats
CO2	Detect and understand the stories within datasets and its applications.
CO3	Apply data visualization best practices
CO4	Design static charts, interactive Dashboards and data stories

Module	Unit	Details	Hrs.	CO
No.	No.			
1	Introd	luction data visualization		
	1.1	Introduction to data visualization and its need.		
		Data analysis lifecycle.		
		10	CO1	
		10	CO2	
		Evolution, From Visual to Story: Bridging the Gap		
	1.2	Data Fundamentals, Collecting data, Preparing Data		
2	Design	n principles and various charts/plots		
	2.1	Gestalt Design principles		
	2.2	Data cleaning using Excel		
	2.3	Introduction to various chart/ plots and its interpretation		
		The Bar Chart		
		The Line Chart	06	CO 3
		The Pie and Donut Charts	00	COS
		The Scatter Plot		
		The Packed Bubble Chart		
		Box plot,		
		The Treemap		
		Plotting of PDF		
3	Storyt	telling and Multivariate displays		
	3.1	Storytelling Multivariate displays		
		The Science of Storytelling		
		The Power of Stories		
		Context in Action		
		Exploratory versus Explanatory Analysis	10	CO4
		Structuring Stories		
		Audience Analysis for Storytelling		
		Steps to Visual Data Storytelling		
		The Important Role of Feedback		
	3.2	Graphical Perception		
	# Mult	tivariate displays		
4	Geosp	atial displays	09	CO4
	4.1	The Heat Map		
		Maps		
		Connecting to Geographic Data		
		Assigning Geographic Roles		
		Creating Geographic Hierarchies		
		Proportional Symbol Maps		
		Choropleth Map		
5	Dashb	poards, interactive and animated displays	10	CO4

5.1	Visual Design Building Blocks		
	Color		
	Stepped Color		
	Reversed Color		
	Color Effects		
	Opacity		
	Mark Borders		
	Mark Halos		
	The Truth about Red and Green		
	Lines		
	Formatting Grid Lines, Zero Lines, and Drop Lines		
	Formatting Borders		
	Formatting, Shading, and Banding		
	Shapes		
	Shape Marks Card		
	Custom Shapes		
	Timelines		
	Bar-in-Bar Charts		
	Likert Visualizations		
	Lollipop Charts		
	Word Clouds		
	Create dashboard		
	Working with dashboard		
	Publishing through dashboard		
5.2	Any other tool for visualization		
#case	study, Publishing dashboard over mobile devices		
	Total	45	

### **Recommended Books:**

Sr. No.	Name/s of Author/s	Title of Book	Name of Publisher with country	Edition and Year of Publication
1.	Sosulski, K.	Data Visualization Made Simple: Insights into Becoming Visual	New York: Routledge.	First edition, 2018
2.	Lindy Ryan	Visual Data Storytelling with Tableau	Pearson Education	First edition, 2018
3.	Kristen Sosulski	Data Visualization made simple	New York: Routledge	First edition, 2019
4.	Cole Nussbaumer Knaflic	Storytelling with Data	Wiley	First edition, 2015

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Course Code	Course Title							
116h54L301			Data V	/isualiz	ation			
	Т	TH P TUT Total						
Teaching Scheme(Hrs.)			02				02	
Credits Assigned		-		01				01
			]	Marks				
Examination	Examination CA		ECE	(D) X X /		D	P&O	Total
Scheme	ISE	IA	ESE	TW	O	P	rau	1 Otal
	-	_	_	25	25			50

### **Term-Work:**

Term work will consist of experiments/ tutorials covering entire syllabus of the course 'Data Visualization'. Students will be graded based on continuous assessment of their term work.

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Course Code		Course Title						
116h54C401		Applied Data Science						
	,	TH P TUT Total					Total	
Teaching Scheme(Hrs.)		03			-			03
Credits Assigned		03						03
				Marks				
Examination	CA		ESE	TW		P	De O	Total
Scheme	ISE	IA	ESE	1 77	O	P	P&O	Total
	30	20	50					100

### Course prerequisites (if any):

Students are expected to have basic knowledge of algorithms and programming experience.

### **Course Objectives**

- To develop understanding of the applied data science in the real world problems.
- To get the understanding of R programming language with respect to data analysis.
- To understand the application of Machine Learning Algorithms for data modeling.
- To apply various data visualization techniques using real-world data sets and analyze the graphs and charts.
- To understand various analytics metrics, processing unstructured text/data, and the ability to investigate hidden patterns.

### **Course Outcomes**

### At the end of successful completion of the course the student will be able to

CO1	Develop an understanding of data science and business analytics.
CO2	Application of Exploratory data analysis (EDA) on Real world problems.
CO3	Understand the basic concept and techniques of Machine Learning regression
	and classification.
CO4	Understand the basic concept and techniques of Machine Learning clustering.

Module	Unit	Details	Hrs.	CO
No.	No.			
1	Introd	10	CO1	
	Proce	10	COI	
	1.1	Introduction to Applied Data Science: What is Data		
		Current landscape of perspectives - Skill sets needed		
		and various application areas. Challenges and skill		
		Sets needed and various applications areas.		
	1.2	Impact of applying Data Science in business scenario,		
		Introduction to need of estimation and validation for		
		added value due to data science		
	1.3	Introduction to the mathematical foundations		
		required for data science. Statistical Inference:		
		Populations and samples, Statistical modeling,		
		Probability distribution, Fitting a model Normal		
		Distribution, Skewness and Kurtosis,		
		Heteroskedasticity, Descriptive Statistics, Higher-Order		
		Moments, Matrices, Maximum-likelihood, Introduction		
		to Brownian Motions, Monte Carlo		
	1.4	Data Scrapping: Introduction, Need, Sources, Web		
		Scrapping, Scrapping of Images, Data Wrangling, ETL		
		Process, Data Munging		
2	Explo	ratory Data Analysis	08	CO2
	2.1	Exploratory Data Analysis and the Data Science Process,		
		Basic tools (plots, graphs and summary statistics) of		
		EDA, Measuring similarity and dissimilarity		
	2.2	Why Preprocessing? Data Cleaning; Data Integration;		
		Data Reduction: Attribute subset selection, Histograms,		
		Clustering and Sampling; Data Transformation & Data		
		Discretization: Normalization, Binning, Histogram		
		Analysis and Concept hierarchy generation.		
	2.3	The Data Science Process: Case Study on Online E-		
		Commerce Dataset		
3		ine Learning Algorithms – Regression	05	CO3
	3.1	Introduction to Machine learning, Linear Regression,		
		Building model, Model Diagnostic		
	3.2	Multiple Linear regression, Logistic Regression,		
4	Mach	ine Learning Algorithms: Classification	12	CO3

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	4.1	Decision Trees , Naive Bayes theorem, Bayes Classifier,						
		Rule-Based Classification, Bayesian Belief networks,						
		CART, Random Forests, k-Nearest Neighbors (k-NN),						
		Hidden Markov Models, Support Vector Machines						
	4.2	Dimensionality Reduction Techniques:						
		Principal Component Analysis, Independent						
		Component Analysis, Singular value decomposition						
	4.3	Model Evaluation and Selection: Metrics, Confusion						
		Matrix, Precision and Recall, Accuracy, False Positives,						
		Techniques to Improve Classification Accuracy						
5	Mach	ine Learning Algorithms: Clustering	10	CO4				
	5.1	Cluster Analysis: Basic Concepts, Partitioning Methods:						
		KMeans, KMediods and hierarchical methods:						
		Agglomerative. Expectation Maximization Algorithm,						
		Radial Basis functions						
	5.2	Cost Function, how to Minimize cost function,						
		coefficients of determination.						
	5.3 Introduction Text processing, plain and simple, Text							
		Classification, Text Summarization.						
		# Self Learning –Mini Project						
	•	Total	45					

# Students should prepare all Self Learning topics on their own. Self-learning topics will enable students to gain extended knowledge of the topic. Assessment of these topics may be included in IA and Laboratory Experiments.

### **Recommended Books:**

Sr.	Name/s of Author/s	Title of Book	Name of	<b>Edition and</b>
No.			<b>Publisher with</b>	Year of
			country	Publication
1.	Han, J., Kamber, M.,	Data mining concepts and	Morgan	2011
	Pei, J.	techniques	Kaufmann	
2.	James, G., Witten,	An introduction	Springer	2013
	D., Hastie, T.,	to statistical		
	Tibshirani, R.	learning with applications		
		in R		
3.	Cathy O'Neil and	Doing Data Science,	O'Reilly	2014
	Rachel Schutt	Straight Talk From The		
		Frontline		
4.	Kevin P. Murphy	Machine Learning: A	ISBN	2013
		Probabilistic Perspective	0262018020	
5.	Mohammed J. Zaki	Data Mining and Analysis:	Cambridg	2014
	and Wagner Miera Jr.	Fundamental Concepts	e	
		and Algorithms	Universit	
			у	
			Press	
6.	Avrim Blum, John	Foundations of Data	ONLINE	2014
	Hopcroft, and	Science		
	RavindranKannan			
7	C R Kothari	Research Mythology	New Age	4 <sup>th</sup> edition
			International	2019
			Publishers	

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<b>Course Code</b>	Course Title								
116h54L401	Applied Data Science								
	7	TH P TUT Total						Total	
Teaching Scheme(Hrs.)	-			02				02	
Credits Assigned		-		0	1			01	
	Marks								
Examination	CA		ECE		0	ъ	P&O	Total	
Scheme	ISE	IA	ESE	TW	J	P	1 &0	1 Otal	
	-	_	_	25	25			50	

### **Term-Work:**

Term work will consist of experiments/ tutorials covering entire syllabus of the course 'Applied Data Science'. Students will be graded based on continuous assessment of their term work.

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<b>Course Code</b>	Course Title								
116h54C501	Data Analytics								
	П	ГН		F	•	,	TUT	Total	
Teaching Scheme(Hrs.)							03		
Credits Assigned		03		_				02	
				Marks					
Examination	CA	CA		TDXX/	0		P&O	Total	
Scheme	ISE	IA	ESE	TW	U	P	1 &0	1 Otal	
	30	20	50					100	

### Course prerequisites (if any):

- Concepts of DBMS
- Data Mining
- Knowledge of basic Machine Learning algorithms and programming language (C/C++/Java/ Python).

### **Course Objectives**

Introduction to the fundamental concepts of Data Analytics through real world case studies Comprehension of data analytics and visualization for decision-making in Geospatial, healthcare and text mining business applications

### **Course Outcomes**

### At the end of successful completion of the course the student will be able to

CO1	Understand basic concepts of data analytics to solve real-world problems
CO2	Apply the data analytics in the field of geospatial system
CO3	Preform the social data analytics
CO4	Preform Time series Analytics and forecasting
CO5	Apply the data analytics in the field of Health care.

Module No.	Unit No.	Details	Hrs.	СО
1	Intro	duction to Data Analytics	05	CO1
	1.1	Introduction to Data Analytics, Different types of data		
		analytics: Descriptive analytics, Diagnostics Analytics,		
		Predictive analytics, Prescriptive analytics		
	1.2	# Self-Learning: LinkedIn analytics, Netflix Analytics,		
		Cricket and FIFA Analytics.		
2		Analytics in GIS	10	_
	2.1	Introduction, Definition of GIS, Evolution of GIS,		
		components of GIS		
	2.2	Vector Data Model : Topology, Non topological		
		Vector models, Attribute Data in GIS, Attribute Data		
		Entry, Vector Data Query, Manipulation of Fields		
		and Attribute Data		
		Raster Data Model: Elements of Raster Data Model,		GOA
		Types of Raster Data, Raster Data Structure, Raster		CO2
		Data Query, Data Compression, Data		CO1
	2.2	Conversion, Integration of Raster and Vector data		
	2.3	Terrain Analysis: Data for Terrain Mapping and		
		analysis, Terrain Mapping, slope and aspect, Surface		
		curvature, Raster vs TIN, View shed and water shed		
	2.4	analysis.		
	2.4	GIS application Case study: A real world problem and		
		its step by step procedure using open source software tools.		
		# Self-learning: QGIS, Hadoop, GeoSpark R		
		PostgreSQL, PostGIS, Python		
3	Grapl	h Analytics	10	
		•		-
	3.1	Introduction to the Social Network, Clustering of		CO3
		Social-Network Graphs, Direct Discovery of		
	2.2	Communities  Portitioning of Crophs Finding Overlanding		-
	3.2	Partitioning of Graphs, Finding Overlapping Communities, Simrank, Counting Triangles,		
		Neighborhood Properties of Graphs		
	3.3	# Self-learning: GraphX tools of Apache.		
1			10	
4	ļ	series Analytics and forecasting	10	-
	4.1	Introduction, Finding and Wrangling Time Series		
		Data, Exploratory Data Analysis for Time Series,		
	4.2	Simulating Time Series Data, Storing Temporal Data,		CO4
	4.2	Statistical Models for Time Series, State Space Models		
		for Time Series, forecasting methods, Testing for		
		randomness, Regression based trend model, random		
		walk model, moving average forecast, exponential		
5	Dete	smoothing forecast, seasonal models analytics in Health Care Systems	10	CO5
3	5.1	Ţ.	10	005
	3.1	Introduction, Components of HER, Benefits of EHR-		

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	Barrier to Adopting HER challenges		
5.2	Mining Sensor Data in Medical Informatics		
	Challenges in Healthcare Data Analysis		
	Sensor Data Mining Applications		
5.3	Natural Language Processing and data mining for		
	clinical text data: Mining Information from Clinical		
	Text, Challenges of Processing Clinical Reports,		
	Clinical Applications		
	Self-learning: Introduction to Social media analytics		
	for healthcare,		
·	Total	45	

<sup>#</sup> Students should prepare all Self Learning topics on their own. Self-learning topics will enable students to gain extended knowledge of the topic. Assessment of these topics may be included in IA and Laboratory Experiments.

### **Recommended Books:**

Sr. No.	Name/s of Author/s	Title of Book	Name of Publisher with country	Edition and Year of Publication
1.	Michael J. de Smith, Michael F. Goodchild and Paul A. Longley	Geospatial Analysis: A Comprehensive Guide to Principles, Techniques, and Software Tools,	Wiley, Second Edition	2019
2.	Anil Maheshwari	Data Analytics	Mc Graw Hill	2017
3.	James, G., Witten, D., Hastie, T., Tibshirani, R.	An introductionto statistical learning with applications in R	Springer	2013
4.	Chandan K. Reddy and Charu C Aggarwal	Healthcare data analytics	Taylor & Francis	2015
5.	Hui Yang and Eva K. Lee	Healthcare Analytics: From Data to Knowledge to Healthcare Improvement	Wiley	2016
6.	Mohammed J. Zaki and Wagner Miera Jr.	Data Mining and Analysis: Fundamental Concepts and Algorithms	Cambridge University Press	2014
7.	U. Dinesh Kumar	Business Analytics	Wiley	2017
8.	Kang-tsung Chang	"Introduction to Geographical Information Systems",	Tata McGraw Hill	Third Edition, 2003
9	Jure Leskovec, Anand Rajaraman, Jeffrey D. Ullman	Mining Massive dataset	Dreamtech Press	2 <sup>nd</sup> edition 2016
10	Albright and Winston	Bussiness Analytics	Cengage Publication	5 <sup>th</sup> edition, 2015
11	Aileen Nielsen	Practical Time Series Analysis	O'Reilly Media, Inc.	1 <sup>st</sup> edition October 2019

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<b>Course Code</b>	Course Title								
116h54L501	Data Analytics								
	Т	TH P TUT Total							
Teaching Scheme(Hrs.)	-			02				02	
Credits Assigned		-		0	1			01	
	Marks								
Examination	CA	CA		TDXX7		ъ	P&O	Total	
Scheme	ISE	IA	ESE	TW	O	P	rao	1 Otal	
	-	_	_	25	25			50	

### **Term-Work:**

Term work will consist of experiments/ tutorials covering entire syllabus of the course 'Data Analytics'. Students will be graded based on continuous assessment of their term work.

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<b>Course Code</b>	Course Title								
116h54C601	Advanced Data mining								
	Г	TH		P	•	,	TUT	Total	
Teaching Scheme(Hrs.)	03							03	
Credits Assigned	(	03						03	
				Marks					
Examination	CA		ESE	TW	0	P	P&O	Total	
Scheme	ISE	IA	ESE	1 44	U	P	rau	1 otai	
	30	20	50					100	

# Course prerequisites: Data mining Course Objectives:

- 1. To analyze various algorithms and techniques to mine complex data beyond conventional record data to mining complex structure and complex data
- 2. Extract useful knowledge from massive data sources using distributed computing solutions for data intensive applications
- 3. To develop the basic skills necessary to pursue research in data mining.

### **Course Outcomes**

On completion of the course students will be expected to

- 1. Describe the fundamental issues and challenges of mining complex data
- 2. Analyze patterns in streaming data
- 3. Derive patterns from complex structures and sequence data
- 4. Understand the concepts of information retrieval and web search
- 5. Analyze patterns in multivariate time series data

Module	Unit	Details	Hrs.	CO			
No.	No.						
1	Data 1	nining Introduction					
	differe	Data mining process, different types of data representation, different types of knowledge mined, common data mining tasks, Distributed computing solution for data mining and applications					
2		nental Datamining and Stream mining					
	Incren Chara	nental algorithms for mining frequent patterns acteristics of Streaming Data, Issues and Challenges, ning Data Mining Algorithms	06	CO2			
3	Minin	g complex structures					
	Mining for mi Applic mining Mining netwo	10	CO3				
4		nining & Web Search					
	Cluste Web s Rank Applic	search: Crawling & Indexing, Hyperlink Analysis, Page algorithm, Web Search and Information Retrieval, eation: Query Recommender System	08	CO4			
5		nce mining & Multivariate and Time series mining					
	Model Applic Multiv and Ti	ince Mining- Characteristics of Sequence Data, Problem ing, Sequential Pattern Discovery, Timing Constraints eations in Bioinformatics variate and Time series mining- Importance of Multivariate me series data, Sources of MVTS data, Mining MVTS Sign Language Data, Agro-meteorological Data	08	CO5			

### **Recommended Books:**

Sr.	Name/s of Author/s	Title of Book	Name of Publisher	<b>Edition and</b>
No.			with country	Year of
				Publication
1.	Hadzic F., Tan H. &	Mining data with	Springer	2011
	Dillon T. S	Complex Structures		
2.	Yates R. B. and Neto	Modern Information	Pearson Education	2005
	B. R	Retrieval		
3.	Han J. & Kamber M	Data Mining: Concepts	Morgan Kaufmann	Third edition,
		and Techniques	Publishers	2011
4.	Christopher D.M.,	Introduction to	Cambridge UP	Online
	Prabhakar R. &	Information Retrieval"		edition,
	Hinrich S			2009
5.	Tan P. N., Steinbach	Introduction to Data	Pearson Education	2006
	M & Kumar V	Mining		

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<b>Course Code</b>	Course Title							
116h54C701	Advanced Machine Learning							
	Г	P		TUT		Total		
Teaching Scheme(Hrs.)	03			2				05
Credits Assigned	(		1				04	
	Marks							
Examination	CA		ESE	TW	0	P	P&O	Total
Scheme	ISE	IA	ESE	1 44		1	1 &0	Total
	30	20	50					100

**Course prerequisites:** Probability concepts, calculus, linear algebra & basic python programming.

### **Course Objectives:**

- 6. To analyse various machine learning algorithms and techniques with a modern outlook focusing on recent advances.
- 7. Explore supervised and unsupervised learning paradigms of machine learning.
- 8. To understand natural language processing and to learn how to apply basic algorithms in this field.
- 9. To develop the basic skills necessary to pursue research in machine learning.

### **Course Outcomes**

On completion of the course students will be expected to

- 1. Describe the fundamental issues and challenges of machine learning: data, model selection, model complexity.
- 2. Understand & describe mathematical foundation behind deep learning architectures.
- 3. Design & implement various deep supervised learning architectures for image data.
- 4. Learn and train various deep learning models in NLP.
- 5. Apply various deep learning techniques to design efficient algorithms for real-world applications.

Module	Unit Details	Hrs.	CO			
No.	No.					
1	Machine learning foundation:					
	What is Machine Learning? Types of learning, applications,					
	Bias, variance, overfitting, underfitting, cross validation and	05	CO1			
	feature engineering, gradient descent learning algorithm and its					
	variations		<u> </u>			
2	Deep learning fundamentals:					
	Mathematical foundations of deep learning & neural network					
	playground. Learning rates and data normalization, activation					
	functions, Optimizers, Regularization, Dropout, Momentum &	10	CO2			
	Batch Norm. Neural Nets-Deep Vs Shallow Networks- deep					
	multi-layer perceptron.					
	#Self-Learning: forward and backward propagation in neural					
	networks.					
3	Deep Learning Architecture:	4				
	Convolutional Neural Networks: Convolution, pooling					
	operations, basic CNN architecture and image classification					
	using CNN.					
	Recurrent Neural Networks: RNN introduction, difference					
	between feed forward and RNN, forward, backward propagation					
	in RNN & Vanishing/Exploding gradient problem.					
	Long Short Term Memory: LSTM gates, forward, backward					
	propagation in LSTM.					
	#Self-Learning: Auto encoders					
4	Deep learning for NLP:	+				
•	NLP introduction, text preprocessing techniques, text	=				
	representation (one hot encoding, bag of N grams, TFIDF					
	features). Word Embedding – word2vec model.	10	CO4			
	Text generation, Seq2Seq models (Encoder decoder					
	architecture).					
	#Self-Learning: NLTK & Gensim library					
5	Machine learning case studies:					
	Churn Analysis and Prediction	]				
	Sentiment Analysis or Topic Mining from New York Times					
	Customer Segmentation and Value	10	CO5			
	Netflix Movie Recommendation System					
	• Image & text classification using TenserFlow & PyTorch.					
	• Transfer learning with Pre-trained CNN using TenserFlow.	<u> </u>				
	Total	45				

### **Recommended Books:**

No.			with country	Year of Publication
1.	Tom M.Mitchell	Machine Learning	McGraw Hill	2017
2.	M. Gopal	Applied Machine Learning	McGraw Hill	2018
3.	Ian Goodfellow, Yoshua Bengio, Aaron Courville	Deep Learning	An MIT Press book	2016
4.	Deng & Yu	Deep Learning: Methods and Applications	Now Publishers	2013
5.	Sowmya Vajjala, Bodhisattwa Majumder, Anuj Gupta and Harshit Surana	Practical Natural Language Processing	O'Reilly Media, Inc.	June 2020
6	Sudharsan Ravichandiran	Hands-On Deep Learning Algorithms with Python	Packt Publishing	July 2019

K. J. Somaiya College of Engineering, Mumbai -77 (A Constituent College of Somaiya Vidyavihar University)

Course Code	Course Title								
116h54P801		Applied Project/Internship							
	TH				P		TUT		Total
Teaching Scheme (Hrs./Week)	-				04		-		04
Credits Assigned		-			02		-		02
	Marks								
Examination	CA			ECE	TXX		D	P&O	Total
Scheme		ISE	IA	ESE	TW	O	P	rau	Total
	-	-	-	-	50	50	-	-	100

Course prerequisites: Conceptual knowledge of Data Science & Analytics

Course Objectives: The objectives are to address a real-world problem, which includes identify and solve the problem by implementing the solution using the courses learned in earlier semesters. Recognize various hardware and software requirements for solving the problem. It will also inculcate qualities such as working in team, meeting deadlines, making and following work plan. The Project may include some software or techniques not covered in the courses taught to provide solution of the chosen problem.

### **Course Outcomes:**

### At the end of successful completion of the course the student will be able to

- CO1. Define the problem statement and scope of problem.
- CO2. Identify various hardware and software requirements for problem solution
- CO3. Describe the design with the help of flowchart/block diagrams or any design Tool.
- CO4. Implement and test the design to meet the desired specifications.
- CO5. Analyze, interpret results and correspondingly modify the designed system to get the desired results.
- CO6. Prepare a technical report and technical paper based on the project.

**Term Work and Oral:** This is an activity to be undertaken by the group of 2 or 3 students. Each group will be assigned one faculty member as a supervisor. There will be continuous assessment of the project and progress report of the project needs to be maintained by students. The final oral will be a presentation based on a demonstration of the project in front of a committee of examiners. Students are expected to publish technical paper based on the project.