NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

1.	Subject Code: AID-52	1		Course Title:	Mathematics for	or Data Science
2.	Contact Hours:	L: 3	T: 1	P :	0	
3.	Examination Duratio	n (Hrs.): T	heory: 3	Practica	d: 0	
4.	Relative Weightage:	CWS: 20-35	PRS: 0	MTE: 20-30	ETE: 40-50	PRE: 0
5.	Credits: 4	6. Semester:	Autumn		7. Subject A	rea: PCC
8.	Pre-requisite: Nil					

9. Objective: To introduce students to the various Mathematical concepts to be used in ML and DS.

10. Details of the Course

S.No.	Contents		
		hours	
1.	Basics of Linear Algebra : Representation of vectors; Linear dependence and independence; vector space and subspaces (definition, examples and concepts of basis); linear transformations; range and null space; matrices associated with linear transformations; special matrices; eigenvalues and eigenvectors with applications to data problems; Least square and minimum normed solutions	10	
2.	Matrices in Machine Learning Algorithms: projection transformation; orthogonal decomposition; singular value decomposition; principal component analysis and linear discriminant analysis	4	
3.	Gradient Calculus: Basic concepts of calculus: partial derivatives, gradient, directional derivatives, Jacobian, Hessian.	4	
4.	Optimization: Convex sets, Convex function and their properties, Unconstrained and Constrained Optimization, Numerical Optimization Techniques for Unconstrained Optimization, Derivative-Free methods (Golden Section, Fibonacci Search Method, Bisecting Method), Methods using Derivatives (Newton's Method, Steepest Descent Method), Penalty Function Methods for Constrained Optimization.	9	
5.	Probability: Basic concepts of probability, conditional probability, total probability, independent events, Bayes' theorem, random variable, Moments, moment generating functions, some useful distributions, Joint distribution, conditional distribution, transformation of random variables, covariance, correlation.	8	
6.	Statistics: Random sample, sampling techniques, statistics, sampling distributions, mixture models.	7	
	Total	42	

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/Reprint
1	M. P. Deisenroth, A. A. Faisal, C. S. Ong, Mathematics for	2020
	Machine Learning, Cambridge University Press (1 st edition)	
2	S. Axler, Linear Algebra Done Right. Springer International Publishing	2015
	(3 rd edition)	
3	J. Nocedal and S. J. Wright, Numerical Optimization. New York:	2006
	Springer Science+Business Media	
4	E. Kreyszig, Advanced Engineering Mathematics, John Wiley and	2015
	Sons, Inc., U.K. (10 th Edition)	
5	R. A. Johnson, I. Miller, and J. E.Freund, "Miller & Freund's	2011
	Probability and Statistics for Engineers", Prentice Hall PTR, (8th	
	edition)	
6	C. Mohan and K. Deep: "Optimization Techniques", New Age	2009
	Publishers, New Delhi.	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

1.	Subject Code: AID-52	23		Course Title:	Principles of Da	tabase Systems
2.	Contact Hours:	L: 3	T: 0	P:	2	
3.	Examination Duration	n (Hrs.):	Theory: 3	Practical	:0	
4.	Relative Weightage:	CWS: 10-25	PRS: 25	MTE: 15-25	ETE: 30-40	PRE: 0
5.	Credits: 4	6. Semeste	er: Autumn		7. Subject Ar	ea: PCC
8.	Pre-requisite: Nil					

9. Objective: To impart the knowledge of basic Data Base Management Systems.

10. Details of the Course

S.No.	Contents	Contact
		hours
1.	Purpose of Database System, Views of data, Data Models, Database Languages- Database System Architecture, Database users and Administrator, Entity Relationship model (E-R model) – E-R Diagrams, Introduction to relational databases.	8
2.	The relational Model – The catalog types, Keys, Relational Algebra, Domain Relational Calculus, Tuple Relational Calculus, Fundamental operations, Additional Operations, SQL fundamentals - Integrity, Triggers, Security, Advanced SQL features, Embedded SQL, Dynamic SQL, Missing Information, Views, Introduction to Distributed Databases and Client/Server Databases.	10
3.	PL/SQL- Basic and Advanced Concepts.	8
4.	Functional Dependencies – Non-loss Decomposition, Functional Dependencies, First, Second, Third Normal Forms, Dependency Preservation, Boyce-Codd Normal Form, Multi-valued Dependencies and Fourth Normal Form – Join Dependencies and Fifth Normal Form.	8
5.	Transaction Concepts - Transaction Recovery, ACID Properties, System Recovery, Media Recovery, Two Phase Commit, Save Points – SQL Facilities for recovery, Concurrency, Need for Concurrency, Locking Protocols, Two Phase Locking, Intent Locking, Deadlock, Serializability – Recovery Isolation Levels – SQL Facilities, for Concurrency.	8
	Total	42

11. List of Practical:

1	DDL and DML commands in SQL-I
2	DDL and DML commands in SQL-II
3	Query designing in SQL
4	Aggregate functions and sorting concepts on created tables
5	Single row operation functions
6	View and displaying data from multiple tables
7	Aggregating data using group functions
8	Designing query with concepts of sub-queries
9	Pl-SQL
10	To implement the concepts of security and privileged
11	Implementing transaction control commands

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/Reprint
1	M. P. Deisenroth, A. A. Faisal, C. S. Ong, Mathematics for	2020
	Machine Learning, Cambridge University Press (1st edition)	
2	S. Axler, Linear Algebra Done Right. Springer International Publishing	2015
	(3 rd edition)	
3	J. Nocedal and S. J. Wright, Numerical Optimization. New York:	2006
	Springer Science+Business Media	
4	E. Kreyszig, Advanced Engineering Mathematics, John Wiley and	2015
	Sons, Inc., U.K. (10 th Edition)	
5	R. A. Johnson, I. Miller, and J. E.Freund, "Miller & Freund's	2011
	Probability and Statistics for Engineers", Prentice Hall PTR, (8th	
	edition)	
6	C. Mohan and K. Deep: "Optimization Techniques", New Age	2009
	Publishers, New Delhi.	

NAME OF DEPARTMENT/CENTRE: Center for Artificial Intelligence and Data Science

- 1. Subject Code: AID-505 **Course Title**: Machine Learning 2. **Contact Hours: L:** 3 **T:** 1 **P:** 0 3. Examination Duration (Hrs.): Theory: 3 **Practical:** 0 Relative Weightage: CWS: 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0 4. Credits: 4 6. Semester: Autumn 7. Subject Area: PCC 5.
- 8. Pre-requisite: Nil
- **9. Objective:** To provide an understanding of the theoretical concepts of machine learning and prepare students for research or industry application of machine learning techniques.

10. Details of the Course

S.No.	Contents	Contact
1.	Introduction: Well-posed learning problems, examples of machine learning applications, model selection and generalization, concept learning, inductive learning hypothesis, inductive bias. Information theory: entropy, mutual information, KL divergence	4
2.	Performance Optimization: Directional Derivatives, Minima, Necessary Conditions for Optimality, Convex function, Gradient Descent, Stable learning rates, Newtons Method, Conjugate gradient method, The Levenberg-Marquardt algorithm.	4
3.	Linear Classification: Linear classifier, Logistic Regression, Decision Boundary, Cost Function Optimization, Multi-class Classification, Bias and Variance, L1 and L2 Regularization, feature reduction, Principal Component Analysis, Singular Value Decomposition	4
4.	Artificial Neural Networks: Perceptron, Linear Networks, Multi-layer Networks, Forward propagation, Backward propagation, Alternative activation functions, variations on backpropagation, Deep neural networks.	5
5.	Decision tree learning: Decision tree representation, appropriate problems for decision tree learning, hypothesis space search in decision tree learning, inductive bias in tree learning, avoiding overfitting the data, alternative measures for selecting attribute values, ensemble methods, bagging, boosting, random forest	5
6.	Support Vector Machines: Computational learning theory, probably approximately correct (PAC) learning, sample complexity and VC dimension, linear SVM, soft margin SVM, kernel functions, nonlinear SVM, Multiclass classification using SVM, Support vector regression.	5
7.	Instance based learning: K-nearest neighbor learning, distance weighted neighbor learning, locally weighted regression, adaptive nearest neighbor methods, The Concept of Unsupervised Learning, Competition networks, K-means clustering algorithm.	3

8.	Bayesian Learning: Bayes theorem, maximum likelihood and least squared error		
	hypotheses, Naive Bayes classifier, Bayesian belief networks, gradient ascent		
	training of Bayesian networks, learning the structure of Bayesian networks, the EM		
	algorithm, mixture of models, Markov models, hidden Markov models.		
9.	Reinforcement learning: the learning task, Q learning, convergence, temporal		
	difference learning, nondeterministic rewards and actions, generalization,	5	
	relationship to dynamic programming.		
Total			

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/Reprint
1.	T. Mitchell, Machine Learning, McGraw Hill	1997
2.	Christopher Bishop, Pattern Recognition and Machine Learning,	2006
	Springer	
3.	K. Murphy. Machine Learning: A probabilistic perspective, MIT Press	2012
4.	Hastie, Tibshirani, Friedman, Elements of statistical learning, Springer	2011
5.	I. Goodfellow, Y. Bengio and A. Courville. Deep Learning. MIT Press	2016
6.	Richard S. Sutton and Andrew G. Barto, Reinforcement Learning: An	2018
	Introduction, MIT Press	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

- 1. Subject Code: AID-525 Course Title: Data Structures and Algorithms
- **2. Contact Hours:** L: 3 T: 1 P: 0
- **3. Examination Duration (Hrs.):** Theory: 3 Practical: 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 46. Semester: Autumn7. Subject Area: PCC
- 8. Pre-requisite: Nil
- 9. Objective: To introduce advanced concepts in data structures and algorithms

10. Details of the Course

S.No.	Contents	Contact hours
1.	Complexity Analysis: Time and Space complexity of algorithms, asymptotic analysis, big O and other notations.	8
2.	Linear Lists, Stacks and Queues, Hashing and Trees: Abstract data types, sequential and linked implementations, equivalence problem, hash table, collision avoidance, linear open addressing, chains, uses of hash tables, insertion, deletion and search operations for sequential and linked lists, doubly linked lists, circular lists, skip lists, applications of lists in bin sort, radix sort, sparse tables, Binary trees and their properties, tree traversal methods and algorithms.	12
3.	Algorithmic Techniques : Algorithm design strategies, divide and conquer, merge sort, quick sort and its performance analysis, randomized quick sort, Strassen's matrix multiplication; Greedy method and its applications, knapsack problem; Dynamic programming and its performance analysis, optimal binary search trees, 0/1 knapsack problem; Traveling salesman problem; Back-tracking, n-queens problem; Branch and bound examples, 15-puzzle problem, 0/1 knapsack, traveling salesman problem.	12
4.	Graph Algorithms : DFS and BFS, spanning trees, bi-connectivity; Minimum cost spanning trees: Kruskal's, Prim's and Sollin's algorithms; Path finding and shortest path algorithms; Topological sorting; Bipartite graphs. P and NP-classes, NP-hard problems, reduction.	10
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication / Reprint
1.	Wirth, N., "Algorithms and Data Structures", Prentice-Hall of India.	2017
2.	Brad Miller and David Ranum, Luther College, "Problem Solving with Algorithms and Data Structures Using Python," Franklin, Beedle &Associates.	2013
3.	Cormen T, Introduction to Algorithms, MIT Press, 3rd Edition.	2009

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

- 1. Subject Code: AID-527 Course Title: Programming for DS
- **2. Contact Hours**: **L**: 0 **T**: 0 **P**: 4
- **3.** Examination Duration (Hrs.): Theory: 0 Practical: 2
- 4. Relative Weightage: CWS: 0 PRS: 50 MTE: 0 ETE: 0 PRE: 50
- 5. Credits: 2 6. Semester: Autumn 7. Subject Area: PCC
- **8. Pre-requisite:** Nil
- **9. Objective:** This course's objective is to provide hands-on experience on the various programming components for Data Science.

10. Details of the Course:

S.No.	Contents	Contact
		hours
1	Python: Basics, Numpy, Pandas, and Matplotlib	16
2	Scikit-Learn and NLTK	12
3	Tensor Flow and Keras	12
4	Tensor Flow Lite: Deploy machine learning systems on mobile (Android	16
	application) (Android Studio, Kotlin/Java)	
	Total	56

S.No.	Name of Authors/Book/Publisher	Year of
		Publication / Reprint
1	Jake VanderPlas "Python Data Science Handbook," First Edition,	2016
1	O'Reilly Media, Inc.	
2	Wes McKinney "Python for Data Analysis: Data Wrangling with	2017
	Pandas, NumPy, and I Python," Second Edition, O'Reilly Media,	
	Inc.	
3	Pramod Singh and Avinash Manure "Learn TensorFlow 2.0:	2020
	Implement Machine Learning and Deep Learning Models with	
	Python," First Edition, Apress	
4	Aurélien Géron "Hands-On Machine Learning with Scikit-Learn,	2019
	Keras, and TensorFlow," Second Edition, O'Reilly Media, Inc.	
5	Bill Phillips "Android Programming: The Big Nerd Ranch Guide,"	2017
	Third Edition, Big Nerd Ranch Guide	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

1.	Subject Code: AID-571			Course	Title : Big Data A	nalytics
2.	Contact Hours:	L: 3	T:	1	P: 0	
3.	Examination Duration	n (Hrs.): Theo	ory: 3	Practic	al: 0	
4.	Relative Weightage:	CWS: 20-35	PRS: 0	MTE: 20-30	ETE: 40-50	PRE: 0
5.	Credits: 4	6. Semester:	Both		7. Subject Ar	ea: PEC

- 8. Pre-requisite: Nil
- **9. Objective:** The purpose of this course is to introduce the students with Big Data Storage Systems and important algorithms that form the basis of Big Data Processing. The course also introduces the students with major application areas of Big Data Analytics.

10. Details of the Course

S.No.	Contents	Contact hours
1.	Introduction to Big Data: Introduction to Big Data, The four dimensions of Big Data: volume, velocity, variety, veracity, Drivers for Big Data, Introducing the Storage, Query Stack, Revisit useful technologies and concepts, Real-time Big Data Analytics	6
2.	Distributed File Systems: Hadoop Distributed File System, Google File System, DataConsistency	6
3.	Big Data Storage Models: Distributed Hash-table, Key-Value Storage Model (Amazon's Dynamo), Document Storage Model (Facebook's Cassandra), Graph storage models	10
4.	Scalable Algorithms: Mining large graphs, with focus on social networks and web graphs. Centrality, similarity, a 11-distances sketches, community detection,1 ink analysis, spectral techniques. Map-reduce, Pig Latin, and NoSQL, Algorithms for detecting similar items, Recommendation systems, Data stream analysis algorithms, Clustering algorithms, Detecting frequent items	10
5.	Big Data Applications: Advertising on the Web, Web Page Quality Ranking, Mining Social-Networking Group, Human Interaction with Big-Data. Recommendation systems with case studies of Amazon's, Item-to-item recommendations and Netflix Prize, Link Analysis with case studies of the PageRankalgorithm and the spam farm analysis, Crowdsourcing Big Data Issues: Privacy, Visualization, Compliance and Security, Structured vs Unstructured Data	6
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication / Reprint
1.	Ohlhorst, Frank J. Big data analytics: turning big data into big	2012
	money. Vol. 65. John Wiley & Sons, 2012.	
2.	Russom, Philip. "Big data analytics." TDWI best practices report,	2011
	fourthquarter 19, no. 4 (2011): 1-34.	
3.	Marr, Bernard. Big Data: Using SMART big data, analytics and	2015
	metrics to make better decisions and improve performance. John	
	Wiley & Sons, 2015.	
4.	LaValle, Steve, Eric Lesser, Rebecca Shockley, Michael S.	2011
	Hopkins, and Nina Kruschwitz. "Big data, analytics and the	
	path from insights to value." MIT sloan management review 52,	
	no. 2 (2011): 21-32.	
5	Leskovec, Jure, Anand Rajaraman, and Jeffrey David Ullman.	2020
	Mining of massive data sets. Cambridge university press, 2020.	

NAME OF DEPARTMENT/CENTRE: Department of Computer Science and Engineering

- 1. Subject Code: CSN-515 Course Title: Data Mining and Warehousing
- **2. Contact Hours:** L: 3 T: 1 P: 0
- **3. Examination Duration (Hrs.):** Theory: 3 Practical: 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 46. Semester: Spring7. Subject Area: PEC
- 8. Pre-requisite: CS-102
- **9. Objective:** To educate students to the various concepts, algorithms and techniques in data mining and warehousing and their applications.

10. Details of the Course

S.No.	Contents	Contact
		nours
1.	Introduction to data mining: Motivation and significance of data mining, data	3
	mining functionalities, interestingness measures, classification of data mining	
	system, major issues in data mining.	
2.	Data pre-processing: Need, data summarization, data cleaning, data integration	6
	and transformation, data reduction techniques —Singular Value Decomposition	
	(SVD), Discrete Fourier Transform (DFT), Discrete Wavelet Transform (DWT),	
	data discretization and concept hierarchy generalization.	
3.	Data warehouse and OLAP technology: Data warehouse definition,	4
	multidimensional data model(s), data warehouse architecture, OLAP server types,	
	data warehouse implementation, on-line analytical processing and mining,	
4.	Data cube computation and data generalization: Efficient methods for data	4
	cube computation, discovery driven exploration of data cubes, complex	
	aggregation, attribute oriented induction for data generalization.	
5	Mining frequent patterns associations and correlations: Basic concepts	6
5.	efficient and scalable frequent item set mining algorithms mining various kinds	0
	of association rules —multilevel and multidimensional association rule mining	
	versus correlation analysis constraint based association mining	
6	Classification and prediction: Definition decision trag induction Bayasian	6
0.	classification rule based classification classification by backpropagation and	0
	support vector machines associative classification lazy learners prediction	
	support vector machines, associative classification, fazy learners, prediction,	
7	Cluster Analysis: Definition Clustering Algorithms partitioning hierarchical	6
7.	density based and hased and model based: Clustering high dimensional data	0
	constraint based eluster englysic outlier englysic density based and distance	
	constraint based cluster analysis, outlier analysis - density based and distance	
0	Date mining on complex date and combinations. Algorithms for mining of	7
δ.	Data mining on complex data and applications: Algorithms for mining of	/
	spatial data, multimedia data, text data: data mining applications, social impacts	
	or data mining, trends in data mining.	42
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/ Reprint
1.	Marakas, George M. Modern data warehousing, mining, and	2003
	visualization: core concepts. Upper Saddle River, NJ: Prentice	
	Hall, 2003.	
2.	Pujari, Arun K. Data mining techniques. Universities press, 2001.	2001
3.	Lee, Mong Li, Hongjun Lu, Tok Wang Ling, and Yee Teng Ko.	1999
	"Cleansing data for mining and warehousing." In International	
	Conference on Database and Expert Systems Applications, pp.	
	751-760. Springer, Berlin, Heidelberg, 1999.	
4.	Wang, John, ed. Encyclopedia of data warehousing and mining. iGi	2005
	Global, 2005.	
5.	Gupta, Gopal K. Introduction to data mining with case studies. PHI	2014
	Learning Pvt. Ltd., 2014.	
6.	Tan, Pang-Ning, Michael Steinbach, and Vipin Kumar.	2016
	Introduction to data mining. Pearson Education India, 2016.	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

- 1. Subject Code: AID-552 Course Title: Deep Learning
- **2.** Contact Hours: L: 3 T: 1 P: 0
- **3.** Examination Duration (Hrs.): Theory: 3 Practical: 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 46. Semester: Both7. Subject Area: PEC
- **8. Pre-requisite:** Machine Learning
- 9. Objective: The objective of this course is to learn deep learning algorithms, concepts, experiments, research along with their application on generic use cases.

10. Details of the Course:

S.No	Contents	Contact Hours
1	Introduction to deep learning, logical computations with neurons, perceptron, backpropagation, historical trends, applications, and use-cases for industry	6
2	Deep Networks: Training a deep neural network (DNN), hidden layers, activation functions, fine-tuning neural network hyper-parameters	7
3	Custom Deep Neural Networks: vanishing/exploding gradient issues, reusing pre trained layers, optimizers, 11 and 12 regularization, dropout	8
4	Convolutional neural networks (CNNs): convolutional layer, filters, stacking, pooling layer, CNN architectures	7
5	Recurrent neural networks (RNNs): recurrent neurons, unrolling, input and output sequences, training RNNs, deep RNNs, LSTM cell, GRU cell	7
6	Representation Learning and Generative Learning: Autoencoders: data representations, linear autoencoder, stacked autoencoders, variational autoencoders	7
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/Reprint
1	Aurélien Géron, "Hands-On Machine Learning with Scikit-Learn,	2019
	Keras, and TensorFlow: Concepts, Tools, and Techniques to Build	
	Intelligent Systems,"Second Edition, O'Reilly Media	
2	Ian Goodfellow, Yoshua Bengio, and Aaron Courville, "Deep	2017
	Learning," First Edition, MIT Press	
3	François Chollet "Deep Learning with Python," First Edition,	2018
	Manning Publication	
4	Rowel Atienza "Advanced Deep Learning with Keras," First Edition,	2018
	Packt Publishing	
5	Sudharsan Ravichandran "Hands-On Deep Learning Algorithms with	2019
	Python," First Edition, Packt Publishing	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

- **1.** Subject Code: AID-572Course Title: Ethics in Data Science
- **2. Contact Hours: L**: 3 **T**: 1 **P**: 0
- **3. Examination Duration(Hrs):** Theory: 3 Practical: 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 46. Semester: Both7. Subject Area: PEC
- 8. Pre-requisite: Nil
- 9. Objective: To introduce the concepts of ethics in data science
- **10. Details of the Course:**

S.No.	Contents	Contact Hours
1	Introduction and Philosophical frameworks for assessing fairness:	6
	Foundations of ethics, early theories of fairness (Utilitarianism etc.);	
	contemporary theories of fairness; significance of ethics in data science; ethics	
	vs. law/compliance/public relations; cultural relativism; "professional" ethics	
	in data science; individuals vs. collectives.	
2	Research Ethics: Data driven research, methods of collection of data;	8
	different types of data: qualitative and quantitative; overview of ethical issues	
	in data-driven organizations; doing ethical data analysis; responsible use of	
	research data; plagiarism; fake data and fabrication of data; creation of data	
	base.	
3	Data ownership, privacy and anonymity: Understanding the difference	8
	between data ownership; data privacy and data anonymity; under- standing the	
	idea behind data surveillance; data privacy vs. data security.	
4	Algorithmic fairness: Discrimination and algorithms; obscure and un-	8
	intentional bias displayed by the algorithms; ethics of data scraping and	
	storage; Mosaic data; found data; and designed data.	
5	Policies on data protection: EU's general data protection rules (GDPR);	8
	digital India policy; personal data protection bill; 2019 ("PDP Bill"); ethical	
	issues on data privacy in context with India, case studies.	
6	Case Studies	4
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/Reprint
1.	Michael J. Quinn "Ethics for the Information Age",	2016
	Seventh Edition, Pearson.	
2.	DJ Patil, Hilary Mason, and Mike Loukides "Ethics and Data	2018
	Science", O'Reilly Media Inc.	
3.	Bill Franks, "97 Things About Ethics Everyone in Data Science	2020
	Should Know", O'Reilly Media Inc.	
4.	Kord Davis, "Ethics of Big Data: Balancing Risk and	2012
	Innovation", O'Reilly Media Inc.	

NAME OF DEPARTMENT/CENTRE: Department of Mathematics

- 1. **Subject Code: MAN-628**
- 2. Contact Hours: **T:**0 **P:** 0 **L:** 3
- 3. Examination Duration (Hrs.): Theory: 3 Practical: 0
- 4. Relative Weightage: CWS: 20-35 PRS: 0 MTE: 20-30 ETE: 40-50 PRE: 0
- 7. Subject Area: PEC 5. Credits: 3 6. Semester: Both
- 8. Pre-requisite: Nil
- 9. Objective: To provide knowledge about basic concepts of Evolutionary Algorithms
- 10. Details of the Course:

S.No.	Contents	Contact Hours
1	Genetic Algorithms: Historical development, GA concepts – encoding, fitness function, population size, selection, crossover and mutation operators, along with the methodologies of applying these operators. Binary GA and their operators, Real Coded GA and their operators	12
2	Particle Swarm Optimization: PSO Model, global best, Local best, velocity update equations, position update equations, velocity clamping, inertia weight, constriction coefficients, synchronous and asynchronous updates, Binary PSO.	10
3	Memetic Algorithms: Concepts of memes, Incorporating local search as memes, single and multi-memes, hybridization with GA and PSO, Generation Gaps, Performance metrics.	5
4	Differential Evolution: DE as modified GA, generation of population, operators and their implementation.	5
5	Artificial Bee Colony: Historical development, types of bees and their role in the optimization process.	5
6	Multi-Objective Optimization: Linear and nonlinear multi-objective problems, convex and non – convex problems, dominance – concepts and properties, Pareto – optimality, Use of Evolutionary Computations to solve multi objective optimization, bi level optimization, Theoretical Foundations	5
	Total	42

- **Course Title:** Evolutionary Algorithms

S.No.	Name of Authors/Book/Publisher	Year of Publication/Reprint
1.	Coello, C. A., Van Veldhuizen, D.A. and Lamont, G.B.: "Evolutionary Algorithms for solving Multi Objective Problems", Kluwer.	2002
2	Deb, K.: "Multi-Objective Optimization using Evolutionary Algorithms", John Wiley and Sons.	2002
3	Deb, K.: "Optimization for Engineering Design Algorithms and Examples", Prentice Hall of India.	1998
4	Gen, M. and Cheng, R.: "Genetic Algorithms and Engineering Design", Wiley, New York.	1997
5	Hart, W.E., Krasnogor, N. and Smith, J.E. : "Recent Advances in Memetic Algorithms". Springer Berlin Heidelberg. New York	2005
6	Michalewicz, Z.: "Genetic Algorithms+Data structures=Evolution Programs", Springer-Verlag, 3rd edition, London, UK.	1992

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

- 1. Subject Code: AID-573 Course Title: Intrusion Detection Systems
- **2. Contact Hours:** L: 3 T: 1 P: 0
- **3. Examination Duration (Hrs.):** Theory: 3 Practical: 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 4 6. Semester: Both 7. Subject Area: PEC
- 8. Pre-requisite: Nil
- 9. Objective: To introduce concepts in intrusion detection systems
- **10. Details of the Course**

S.No.	Contents	Contact hours
1.	Introduction to IDS: Intruder types, intrusion methods, processes and detection, message integrity and authentication, honey pots	8
2.	IDS Models: General IDS model and taxonomy, data mining based IDS, Denning model, Framework for constructing features, and different models for intrusion detection systems, SVM, probabilistic, and statistical modelling, evaluation of IDS, cost sensitive IDS	8
3.	Network Security Threat Detection: NBAD, specification based and rate based DDOS, scans/probes, predicting attacks, network based anomaly detection, stealthy surveillance detection; defending against DOS attacks in scout, signature-based solutions, snort rules	9
4.	Host based Threat Detection: Host-based anomaly detection, taxonomy of security flaws in software, self-modelling system calls for intrusion detection with dynamic window size	9
5.	Secure Intrusion Detection Systems: Network security, secure intrusion detection environment, secure policy manager, secure IDS sensor, alarm management, intrusion detection system signatures, sensor configuration, signature and intrusion detection configuration, IP blocking configuration, intrusion detection system architecture.	8
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication / Reprint
1.	J. Paul Guyer, "An Introduction to Intrusion Detection Systems,"	2017
	Create space Independent Publishers	
2.	Gerard Blokdyk, "Intrusion-detection System: How-to," Create space	2017
	Independent Publishers.	
3.	Rash, M., Orebaugh, A. and Clark, G., "Intrusion Prevention and	2005
	Active Response: Deploying Network and Host IPS," Syngress.	
4.	Endorf, C., Schultz E. and Mellander J., "Intrusion Detection and	2003
	Prevention," McGraw-Hill.	

NAME OF DEPARTMENT/CENTRE: Department of Computer Science and Engineering

- 1. Subject Code: CSN-528 **Course Title**: Natural Language Processing **P:** 0 2. **Contact Hours: L:** 3 **T:** 0 3. Examination Duration (Hrs.): **Practical:** 0 **Theory:** 3 4. Relative Weightage: CWS: 20-35 **MTE:** 20-30 **PRS:** 0 **ETE:** 40-50 **PRE:** 0 5. Credits: 3 6. Semester: Both 7. Subject Area: PEC
- 8. Pre-requisite: Basic knowledge of Artificial Intelligence
- **9. Objective:** To provide an understanding of the theoretical concepts of Natural Language Processing and prepare students for research or industry application of Natural Language Processing.

10. Details of the Course

S.No.	Contents	
		hours
1.	Introduction to NLP, Corpus, Representation of Words, Preprocessing, Linguistic	6
	and Statistical Properties of Words, POS Tagging, Parsing, Performance	
	Measures, Error Analysis, Confusion Matrix	
2.	Probability and NLP, n-Gram, Language Model, Join and Conditional	6
	Probability, Chain Rule, Markov Assumption, Data Sparsity, Smoothing	
	Techniques, Generative Models, Naive Bayes	
3.	Distributed representation of words for NLP, Co-occurrence Matrix,	6
	Collocations, Dimensionality Reduction, Singular Value Decomposition	
4.	Document Similarity, Inverted Index, Word2Vec, C-BoW, Skip-Gram Model,	6
	Sampling, Hierarchical Soft-max, Sequence Learning	
5.	Neural Networks for NLP, Multi-Layer Perceptron, Activation Function,	6
	Gradient Descent, Sequence Modeling, Recurrent Neural Networks	
6.	Gated Recurrent Unit, Long-Short Term Memory Networks, 1-D Convolutional	6
	Layer, Language Model using RNN, Forward Pass, Backward Pass	
7.	Applications of NLP, Topic Modeling, Sentiment Analysis, Query Processing,	6
	ChatBoat, Machine Translation, Statistical Machine Translation, Neural Machine	
	Translation, Spell Checker, Summarization	
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication / Reprint
1.	Manning, Christopher, and Hinrich Schutze. Foundations of statistical natural language processing. MIT press, 1999.	1999
2.	Jurafsky, Dan. Speech & language processing. Pearson Education India, 2000.	2000
3.	Smith, Noah A. Linguistic structure prediction. Morgan and Claypool, 2011.	2011
4.	Kennedy, Graeme. An introduction to corpus linguistics. Routledge, 2014.	2014

NAME OF DEPARTMENT/CENTRE: Department of Mathematics

- 1. Subject Code: MAN-613 Course Title: Operations Research 2. **Contact Hours: L:** 3 **T:** 1 **P:** 0 **Theory:** 3 3. **Examination Duration (Hrs.):** Practical: 0 **MTE:** 20-30 4. **Relative Weightage:** CWS: 20-35 **PRS:** 0 **ETE:** 40-50 **PRE:** 0 Credits: 4 7. Subject Area: PEC 5. 6. Semester: Both
- 8. Pre-requisite: Nil
- 9. Objective: To acquaint the students with the basic techniques of Operations Research.
- **10. Details of the Course:**

S.No.	. Contents	
		Hours
1	Basics of LPP: Different Types of OR Models, Convex Sets, Graphical Method,	11
	Simplex Method, Big -M Method, Two Phase Method, Revised Simplex	
	Method.	
2	Duality Theory: Dual Simplex Method, Sensitivity Analysis, Parametric Linear	9
	Programming.	
3	Integer Program: Cutting Plane and Branch and Bound Techniques for all Integer	5
	and Mixed Integer Programming Problems	
4	Transportation Problems: Transportation Problems and Assignment	5
	Problems.	
5	Game Theory: Graphical Method and Linear Programming Method for	5
	Rectangular Games, Saddle point, notion of dominance.	
6	Queuing Theory: Steady -state solutions of Markovian Queuing Models: M/M/1,	7
	M/M/1 with limited waiting space, M/M/C, M/M/C with limited space, M/G/1,	
	Inventory Models.	
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/Reprint
1.	Mohan, C. and Deep, K.: "Optimization Techniques", New Age	2009
	India Pvt. Ltd, New Delhi.	
	Mittal, K.V. and Mohan, C.: "Optimization Methods in System	
2.	Analysis and Operations Research", New Age India Pvt. Ltd, New	1996
	Delhi.	
3.	Taha, H.A.: "Operations Research: An Introduction", MacMillan	2013
	Pub Co., NY, Ninth Edition (Reprint).	
	Ravindran, A., Phillips, D.T. and Solberg, J.J.: "Operations	
4	Research: Principles and Practice", John Wiley and Sons, NY,	2012
	Second Edition (Reprint).	
5	Pant, J.C.: "Introduction to Optimization/ Operations Research",	2012
	Jain Brothers, New Delhi, Second Edition.	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence & Data Science

- 1. Subject Code: AID-554 Course Title: Reinforcement Learning
- **2.** Contact Hours: L: 3 T: 1 P: 0
- **3. Examination Duration (Hrs.):** Theory: 3 Practical: 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 46. Semester: Both7. Subject Area: PEC
- 8. Pre-requisite: Nil
- **9. Objective:** This course aims to understand several reinforcement learning algorithms and their applications, along with emerging research trends.

10. Details of the Course:

S.No.	Contents	
		Hours
1	Basics of probability and linear algebra, Definition of a stochastic multi-armed	6
	bandit, Definition of regret, Achieving sublinear regret, UCB algorithm, KL-UCB,	
	Thompson Sampling.	
2	Markov Decision Problem, policy, and value function, Reward models (infinite	8
	discounted, total, finite horizon, and average), Episodic & continuing tasks,	
	Bellman's optimality operator, and Value iteration & policy iteration	
3	The Reinforcement Learning problem, prediction and control problems, Model-	8
	based algorithm, Monte Carlo methods for prediction, and Online implementation	
	of Monte Carlo policy evaluation	
4	Bootstrapping; TD(0) algorithm; Convergence of Monte Carlo and batch	6
	TD(0) algorithms; Model-free control: Q-learning, Sarsa, Expected Sarsa.	
5	n-step returns; $TD(\lambda)$ algorithm; Need for generalization in practice; Linear	6
	function approximation and geometric view; Linear $TD(\lambda)$.	
6	Tile coding; Control with function approximation; Policy search; Policy	8
	gradient methods; Experience replay; Fitted Q Iteration; Case studies.	
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/ Reprint
1	Sutton, Richard S., and Andrew G. Barto. "Reinforcement learning: An introduction," First Edition, MIT press	2020
2	Sugiyama, Masashi. "Statistical reinforcement learning: modern	2015
	machine learning approaches," First Edition, CRC Press	

3	Lattimore, T. and C. Szepesvári. "Bandit algorithms," First Edition, Cambridge University Press.	2020
4	Boris Belousov, Hany Abdulsamad, Pascal Klink, Simone Parisi, and Jan Peters "Reinforcement Learning Algorithms: Analysis and Applications,"First Edition, Springer	2021
5	Alexander Zai and Brandon Brown "Deep Reinforcement Learning in Action," First Edition, Manning Publications	2020

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

- 1. Subject Code: AID-574 Course Title: Spreadsheet Modeling and Simulation
- **2.** Contact Hours: L: 3 T: 1 P: 0
- **3. Examination Duration (Hrs.): Theory:** 3 **Practical:** 0
- 4. Relative Weightage: CWS: 20-35 PRS: 0 MTE: 20-30 ETE: 40-50 PRE: 0
- 5. Credits: 46. Semester: Both7. Subject Area: PEC
- 8. Pre-requisite: Nil
- 9. Objective: To introduce the concept of spreadsheet modelling and simulations to the students.

10. Details of the Course:

S.No.	Contents	Contact Hours
1	Introduction: Introduction to spreadsheets; historical development; basic	6
	capabilities of spreadsheets and their usage for creating models; types of	
	data used in spreadsheets; spreadsheet notations for mathematical	
	operations; common built-in formulas and functions; conditional	
	expressions; relative and absolute references.	
2	Model building: Designing spreadsheets reflecting assumptions; decision	10
	variables; and outcomes, creating basic cash-flow models; revaluating	
	small business opportunities; incorporating what-if analysis; identifying	
	key variables using sensitivity analysis; linear programming models and	
	deterministic models.	
3	Optimization with Spreadsheets using Solver: Linear programming,	12
	sensitivity analysis, transportation and assignment problems, network	
	optimization problems, integer and nonlinear programming, multi-objective	
	optimization, applications of optimization in different areas.	
4	Simulation and Optimization: Use of spreadsheets to implement Monte	10
	Carlo simulations and linear programs for optimization; model uncertainty	
	and risk in spreadsheets; and use of Excel's solver.	
5	Case Studies	4
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of Publication/Reprint
1.	Hillier and Hillier "Introduction to Management Science:	2013
	A Modeling and Case Studies Approach with	
	Spreadsheets", McGraw-Hill/Irwin.	
2.	Cliff Ragsdale "Spreadsheet Modeling and Decision Analysis:	2018
	A Practical Introduction to Business Analytics", Cengage India.	
3.	Barry Render, Nagraj Balakrishnan, and Ralph Stair,	2004
	"Managerial Decision Modelling with Spreadsheets", Pearson.	
4.	S. Christian Albright and Wayne Winston "Spreadsheet	2004
	Modeling and Applications: Essentials of Practical	
	Management Science", Cengage.	

NAME OF DEPARTMENT/CENTRE: Department of Mathematics

- 1. Subject Code: MAN-526Course Title: Soft Computing
- **2.** Contact Hours: L: 3 **T**: 0 **P**: 0
- **3.** Examination Duration (Hrs.): Theory: 3 Practical: 0
- 4. Relative Weightage: CWS: 20-35 PRS: 0 MTE: 20-30 ETE: 40-50 PRE: 0
- 5. Credits: 36. Semester: Both7. Subject Area: PEC
- 8. **Pre-requisite:** Nil
- 9. Objective: To acquaint the students with the basic techniques of Soft Computing.
- **10. Details of the Course:**

S.No.	Contents	Contact		
		Hours		
1	Introduction to Soft Computing, Historical Development, Definitions,	2		
	advantages and disadvantages, solution of complex real life problems			
2	Neural Networks: Fundamentals, Neural Network Architectures, Feedforward	10		
	Networks, Backpropagation Networks.			
3	Fuzzy Logic: Fuzzy Sets, Fuzzy numbers, Fuzzy Systems, membership	8		
	functions, fuzzification, defuzzification.			
4	Genetic Algorithms: Generation of population, Encoding, Fitness Function,	10		
	Reproduction, Crossover, Mutation, probability of crossover and probability			
	of mutation, convergence.			
5	Hybrid Systems: Genetic Algorithm based Backpropagation Network, Fuzzy-	7		
	Backpropagation, Fuzzy Logic Controlled Genetic Algorithms. Case studies.			
6	Case studies in Engineering	5		
	Total	42		

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/Reprint
1.	Jang, J-S. R., Sun,C-T, Mizutani, E.: "Neuro-Fuzzy and Soft	2002
	Computing", Prentice Hall of India.	
2.	Klir, G. J. and Yuan, B.: "Fuzzy Sets and Fuzzy Logic: Theory and	1995
	Applications", Prentice Hall.	1775
3.	Rajasekaran, S. and Vijayalakshmi Pai, G.A.: "Neural Networks,	2003
	Fuzzy Logic and Genetic Algorithms: Synthesis and Applications",	
	Prentice Hall of India.	
4	Sinha, N.K. and Gupta, M. M. : "Soft Computing and Intelligent	2000
	Systems - Theory and Applications", Academic Press.	
5	Tettamanzi, A., Tomassini, M.: "Soft Computing: Integrating	2001
	Evolutionary, Neural, and Fuzzy Systems", Springer.	

NAME OF DEPARTMENT/CENTRE: Department of Mathematics

- 1. Subject Code: MAN-507Course Title: Statistical Inference
- **2. Contact Hours:** L: 3 T: 0 P: 0
- **3.** Examination Duration (Hrs.): Theory: 3 Practical: 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 36. Semester: Both7. Subject Area: PEC
- 8. Pre-requisite: Nil
- 9. Objective: To introduce the concepts of statistical inference.
- **10. Details of the Course**

S.No.	Contents			
		hours		
1.	Principle of Data Reduction: Sufficiency principle, Factorization criterion,	8		
	minimal sufficiency, Completeness and bounded completeness, Likelihood			
	principle, Equivariance principle.			
2.	Theory of Estimation: Basic concepts of estimation, Point estimation, methods of	12		
	estimation; method of moments, method of maximum likelihood; Unbiasedness,			
	Minimum variance estimation, Cramer - Rao bound and its generalization, Rao			
	Blackwell theorem, Existence of UMVUE estimators. Interval Estimation, Some			
	results for normal population case.			
3.	Testing of Hypothesis: Null and alternative hypothesis, Type I and II errors error	18		
	probability and power function, Method of finding tests, Neyman – Pearson lemma,			
	Uniformly most powerful tests, Likelihood ratio principle, Likelihood ratio test,			
	Sequential probability ratio test, Some results based on normal population.			
4.	Analysis of Variance: one-way classification; simple linear regression analysis	4		
	with normal distribution.			
	Total	42		

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/ Reprint
1.	Miller, I. and Miller, M., "Freund's Mathematical Statistics with	2006
	Applications", Prentice Hall PTR, 7th edition	
2.	Lehman, E.L., "Testing of Statistical Hypothesis", Wiley Eastern Ltd	1959
3.	G. Casella, R. L. Berger, "Statistical Inference", Duxbury Press	2002
4.	Lehman, E.L., "Point Estimation", John Wiley & sons	1984
5.	Rohatgi, V.K., "Statistical Inference", Dover Publications	2011

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

1. Subject Code: AID-555 Course Title: Time Series Data Analysis

2.	Contact Hours :	L: 3	T:	l	P: 0	
3.	Examination Duration	on (Hrs.):	Th	eory: 3	Practical: 0	
4.	Relative Weightage:	CWS: 20-35	PRS: 0	MTE: 20-30	ETE: 40-50	PRE: 0
5.	Credits: 4	6. Semester:	Both	7. Subject Are	a: PEC	

- 8. Pre-requisite: Nil
- **9. Objective:** The objective of this course is to understand and analyze time-series data facilitated by R programming

10. Details of the Course:

S.No.	Contents	
		Hours
1	Basic Properties of time-series data: Distribution and moments, Stationarity,	4
	Autocorrelation, Heteroscedasticity, Normality	
2	Autoregressive models and forecasting: AR, ARMA, ARIMA models	4
3	Random walk model: Non-stationarity and unit-root process, Drift and Trend	4
	models	
4	Regression analysis with time-series data using R programming	5
5	Principal Component Analysis (PCA) and Factor Analysis	5
6	Conditional Heteroscedastic Models: ARCH, GARCH. T-GARCH, BEKK-	6
	GARCH	
7	Introduction to Non-linear and regime-switching models: Markov regime-	5
	switching models, Quantile regression, Contagion models	
8	Introduction to Vector Auto-regressive (VAR) models: Impulse Response	5
	Function (IRF), Error Correction Models, Co-integration	
9	Introduction to Panel data models: Fixed-Effect and Random-Effect models	4
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication /Reprint
1	Chris Brooks "Introductory Econometrics for Finance," Fourth	2019
-	Edition, Cambridge University Press	2017
2	Ruey S. Tsay "Analysis of Time-series data," Third Edition, Wiley	2014
3	John Fox and Sanford Weisberg "An R Companion to Applied	2018
	Regression," Third Edition, SAGE	
4	Yves Croissant and Giovanni Millo "Panel Data Econometrics with	2018
	R," First Edition, Wiley	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

1.	Subject Code: AID-57	75		Course 7	fitle : Blockchai	n Technology
2.	Contact Hours:	L: 3	T: 1	P :	0	
3.	Examination Duration	n (Hrs.): Th	eory: 3	Practica	d: 0	
4.	Relative Weightage:	CWS: 20-35	PRS: 0	MTE: 20-30	ETE: 40-50	PRE: 0
5.	Credits: 4	6. Semest	er: Both		7. Subject A	Area: PEC

- 8. Pre-requisite: Nil
- **9. Objective:** To provide knowledge about cryptographic and cybersecurity concepts of blockchain technology with some applications.

10. Details of the Course

S.No.	Contents	Contact hours
1.	Basics of Blockchain: Distributed Database, Two General Problem, Byzantine	5
	General problem and Fault Tolerance, Hadoop Distributed File System, Distributed	
	Hash Table, ASIC resistance, Turing Complete	
2.	Crypto Primitives: Hash functions, security aspects of hash function, Collison	8
	resistant hash, digital signatures, public key cryptography, verifiable random	
	functions, NIST standards	
3.	Blockchain Theory: Advantage over conventional distributed database,	8
	Blockchain Network, Mining Mechanism, Distributed Consensus, Merkle Patricia	
	Tree, Gas Limit, Transactions and Fee, Anonymity, Reward, Chain Policy, Life of	
	Blockchain application, Soft & Hard Fork, Private and Public blockchain.	
3.	Distributed Consensus: Nakamoto consensus, Proof of Work, Proof of Stake,	6
	Proof of Burn, Difficulty Level, Sybil Attack, Energy utilization and alternate	
4.	Cryptocurrency and regulations: Distributed Ledger, Bitcoin protocols - Mining	10
	strategy and rewards, Ethereum - Construction, DAO, Smart Contract, GHOST,	
	Vulnerability, Attacks, Sidechain, Namecoin, IBM hyper ledger, Stakeholders,	
	Roots of Bitcoin, Legal Aspects - Cryptocurrency Exchange, Black Market and	
	Global Economy	
5.	Blockchain Applications: Good blockchain examples and how to identify	5
	potential use-cases, Design Thinking, Internet of Things, Medical Record	
	Management System, Domain Name Service	
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication / Reprint
1.	Arvind Narayanan, Joseph Bonneau, Edward Felten, Andrew Miller	2016
	and Steven Goldfeder, Bitcoin and Cryptocurrency Technologies: A	
	Comprehensive Introduction, Princeton University Press	
2.	Bettina Warburg, Bill Wanger, Tom Serres, "Basics of Blockchain"	2019
	Independently published	
3.	Andreas M. Antonopoulos, "Mastering Bitcoin: unlocking digital	2014
	cryptocurrencies", O'Reilly Media Inc.,	
4.	Wattenhofer, Roger, "Blockchain Science", Inverted Forest	2019
	Publishing, 3 rd Edition	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

- **1. Subject Code**: AID-568 **Course Title**: ML and AI Applications in Earth Sciences
- **2. Contact Hours**: L: 3 T: 1 P: 0
- **3. Examination Duration (Hrs): Theory:** 3 **Practical:** 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 4 6. Semester: Both 7. Subject Area: PEC
- **8. Pre-requisite:** Good foundation in Mathematics and Physics with specific exposure in Numerical Methods. Understanding of fundamental principles of Geology and Geophysics would be preferable.
- **9. Objective:** To make the participants familiar with tools and techniques in Earth Sciences and the use of Machine Learning and Artificial Intelligence for optimizing the workflows for more accurate prediction of events and properties of the subsurface.

10. Details of the Course:

S.No.	Contents	
		Hours
1	Familiarization with Major Domains and Data Types in Earth Sciences:	4
	Earthquake Seismology, Engineering Geology and Rock Mechanics, Reservoir	
	Characterization, Paleontology	
2	General Introduction to ML and AI in Earth Sciences:	6
	ML and statistical pattern recognition: Supervised learning (generative/ descriptive	
	learning, parametric/ non-parametric learning, neural networks, Support vector	
	machines), Unsupervised learning (clustering, dimensionality reduction, kernel	
	methods); time series modelling, linear regression, regularization, linear classifiers,	
	ensemble methods, neural networks, model selection and evaluation, scalable	
	algorithms for big data, and data ethics.	
	Data science: Extreme value statistics, multi-variate analysis, factor analysis,	
	compositional data analysis, spatial information aggregation models, spatial estimation,	
	geo-statistical simulation, treating data of different scales of observation, spatio-	
	temporal modelling (geo-statistics).	
3	Automating Data Mining and Analysis in Seismology: Basics of earthquake	6
	detection and phase picking using short-term average (STA)/long-term average (LTA);	
	detection using waveform similarity: Network Matched Filtering/template matching,	
	Fingerprint And Similarity Thresholding (FAST). Associating seismic phases across all	
	stations using deep-learning techniques and combining the ones have the same origin	
	source (PhaseLink). Generic workflow of data collection, preprocessing, model	
	training, model evaluation, and production. Applications of ML in ground motion	
	synthesis, and future directions.	
4	Classification of Earthquake Sources: Using supervised learning for classifying	4
	earthquakes and finding their occurrence mechanism. Training dataset (waveforms) on	
	different kinds of sources: earthquake, glacial, volcanic, landslide, explosion, etc. A	
	brief discussion on seismic sources and radiation pattern of emerging waves.	
5	Deep learning (DL) based Seismic Inversion: Theory of Seismic Inversion,	4
	Convolutional neural network (CNN) and fully connected network (FCN) architectures,	
	Performance evaluation, Geophysical inversion versus ML, their applications to	
	reflectivity inversion in seismic, Numerical examples.	

6	Automation in 3D Reservoir Property Prediction: Data Mining, Automated	4
	Petrophysics, Statistical and Regression Methods for Elastic Property Prediction, ML	
	and AI application in Geostatistics, Convoluted Neural Networks for Seismic	
	Interpretation, Deep Learning for Impedance Inversion and Porosity Prediction.	
7	Data-Driven Analytics in Shale Resources: Concepts of shale as source-reservoir-	4
	seal, Modeling Production from Shale, Shale Analytics, Decline Curve Analysis, Shale	
	Production Optimization Technology (SPOT), Numerical Simulation and Smart Proxy	
8	Machine learning Applications in Engineering Geology and Rock Mechanics: ML	6
	in rock mass characterization, Rock Mass Rating, Slope Mass Rating, Q-System,	
	Engineering properties of rock and various rock engineering applications, AI in	
	Landslides study.	
9	Separation and Taxonomic Identification of Microfossil: 3D object recognition and	4
	segmentation applied to X-ray MicroCT images; Testing different algorithms for	
	identifying and localizing individual microfossils in rock samples: Automated	
	Computer Vision, Deep learning-based CNN semantic, and other segmentation	
	architectures.	
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publications/ Reprint
1	Patrick Wong, Fred Aminzadeh, and Masoud Nikravesh, Soft	2002
	Computing for Reservoir Characterization and Modeling, Springer-	
	Verlag Berlin Heidelberg GmbH	
2	William Sandham & Miles Leggett, Geophysical Applications of	2003
	Artificial Neural Network and Fuzzy Logic, Springer	
3	C. Cranganu, H. Luchian, M. E. Breaban, Artificial Intelligent	2015
	Approached in Petroleum Geosciences, Springer	
4	Shahab D. Mohaghegh, Data-Driven Analytics in Unconventional	2017
	Resources, Springer	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

1.	Subject Code: AID-57	Course T	itle: Data Science	ce in Bioinforma	tics	
2.	Contact Hours:	L: 3	T: 1	P :	0	
3.	Examination Duration	n (Hrs.): Th	eory: 3	Practica	l: 0	
4.	Relative Weightage:	CWS: 20-35	PRS: 0	MTE: 20-30	ETE: 40-50	PRE: 0
5.	Credits: 4	6. Semester:	Both	7. Sub	ject Area: PEC	

- 8. Pre-requisite: Nil
- **9. Objective:** The course provides exposure to the Data Science within the context of its importance in biology. The course discusses various methodologies and techniques as well as relevant problems in biology addressed using Data Science.

10. Details of the Course

S.No.	Contents	Contact
		hours
1.	Relevance of Data Science in Bioinformatics. Why Data Science in Biology and	7
	Healthcare? Visualization tools for biological and bioinformatics datasets; data	
	handling; transformations of data	
2.	Data Science in genomics. From genetics to genomes. Alignment and	7
	phylogenetic trees.	
3.	Structural bioinformatics, Proteomics, Protein structure prediction, integrative	7
	structural modeling, and structure-based drug design.	
4.	AI algorithms, statistical tools, graph algorithms for bioinformatics data analytics	7
5.	Deep learning algorithms in perspective of bioinformatics applications; GANs	7
	for biological applications	
6.	Whole-cell modeling approaches, Big Data Consortiums; Hands-on experience	7
	of applying Data Science in Biology	
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication / Reprint
1.	Arthur M. Lesk, "Introduction to Bioinformatics", Oxford University	2019
	Press) (Fifth Edition)	
2.	Jeil Grus, "Data Science from Scratch: First Principles with Python",	2019
	O'Reilly Media Inc. (Second Edition,)	
3.	Vince Buffalo, "Bioinformatics Data skills", O'Reilly Media Inc.	2015
4.	Neil C. Jones and Pavel A. Pevzner, "An introduction to	2004
	Bioinformatics Algorithms", The MIT Press	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

1.	Subject Code: AID-577			rse Title: Data So	cience for Decision	n Making
2.	Contact Hours:	L: 3	T: 1	Р:	0	
3.	Examination Duratio	n (Hrs.): Tl	heory: 3	Practical	: 0	
4.	Relative Weightage:	C WS: 20-35	PRS: 0	MTE: 20-30	ETE: 40-50	PRE: 0
5.	Credits: 4	6. Semes	ster: Both	7. S	ubject Area: PEC	

- 8. Pre-requisite: Nil
- 9. Objective: To introduce the concept of data driven decision making systems to the students.

10. Details of the Course:

S.No.	Contents	Contact Hours
1	Fundamentals of Analytics: Introduction to data-driven decision	8
	making; general introduction to data driven strategy and its importance;	
	use of examples and mini-case studies to illustrate the role of statistical	
	analysis in decision making.	
2	Basic Data Analysis: Various types of data that are commonly collected	10
	by firms; methods to be used and inferences/insights that can be	
	obtained depending on the type of data that are available (stated versus	
	revealed preference, level of aggregation, cross- sectional, time series,	
	comparisons and cross tabulation; statistical informaces using chi	
	comparisons, and cross tabulation, statistical interences using chi-	
3	Experimental Design and Natural Experiments: Issues of design of	10
5	experiments and internal and external validity: case studies in marketing:	10
	economics: and medicine etc.: A-B testing: and circumstances that	
	provide us with "natural" experiments.	
4	Decision making tools: Regression analysis and its applications; use of	10
	regression output in forecasting; promotional planning and optimal pricing;	
	multivariate analysis (unsupervised learning) cluster analysis; factor	
	analysis decision trees; elastic nets and random forests.	
5	Case Studies: To understand the problem at an intuitive level; use of simple	4
	data analysis and visualization to verify (or falsify) the intuition; use of	
	appropriate statistical analysis to present your arguments.	
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/Reprint
1.	F.S. Hillier and G.J. Liberman "Introduction to	2001
	Operations Research" Tata McGraw Hill Education Private	
	Limited.	
2.	Gregory S. Parnel, Terry A. Bresnick, Steven N. Tani, Eric	2013
	R.Johnson "Handbook of Decision Analysis", Wiley.	
3.	Emily Moberg and Igor Linkov "Multi-Criteria Decision	2011
	Analysis: Environmental Applications and Case Studies",	
	CRCPress, Taylor and Francis group.	
4.	Adiel Teixeira de Almeida, Emel Aktas, Sarah Ben Amor,	2020
	João Luis de Miranda "Advanced Studies in Multi-Criteria	
	Decision Making", CRC Press.	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

1.	Subject Code: AID-562Cou			urse Title: AI for Investment		
2.	Contact Hours :	L: 3	T: 0	P: 2		
3.	Examination Dura	tion (Hrs.):	Theory: 3	Practical:	0	
4.	Relative Weightag	e: CWS: 10-25	5 PRS: 25	MTE: 15-25	ETE: 30-40	PRE: 0
5.	Credits: 4	6. Semester: B	oth	7. Subject Area	a: PEC	

- 8. Pre-requisite: Nil
- **9. Objective:** The objective of this course is to understand the application of Artificial Intelligence and Machine Learning techniques in financial markets, trading, and asset management.

10. Details of the Course:

S.No.	Contents		
		Hours	
1	Introduction to financial markets and market microstructure	4	
2	Introduction to risk-return framework	4	
3	Introduction to asset management and portfolio optimization	4	
4	Market efficiency and behavioral finance	4	
5	Prediction in Financial markets using AI and machine learning models, AI and	6	
	machine learning in Trading execution and portfolio management		
6	Credit scoring and credit modeling with non-linear machine learning models	4	
	and deep learning		
7	Model risk management and stress testing	4	
8	Robo advisory, social and quantitative investing	5	
9	Machine learning for asset management	4	
10	AI and machine learning in regulatory compliance and supervision	3	
	Total	42	

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/Reprint
1	M. Dixon, I Halperin, and P. Bilokon "Machine Learning in Finance," First Edition, Springer	2020
2	Marcos Lopez "Advances in Financial Machine Learning," First	2018
	Edition, Wiley	
3	Marcos Lopez "Machine Learning for Asset Managers," First	2020
	Edition, Cambridge University Press	
4	Stefan Jansen "Machine Learning for Algorithmic Trading," Second	2020
	Edition, Packt	2020
5	Elton and Gruber, "Modern Portfolio Theory," Ninth Edition, Wiley	2014

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

1.	Subject Code: AI	(Course Title: Digital Image Processing			
2.	Contact Hours:	L: 3	T: 1		P: 0	
3.	Examination Duration (Hrs.):		Theory: 3	Practical: 0		
4.	Relative Weight:	CWS: 20-35	PRS: 0	MTE: 20-30	ETE: 40-50	PRE: 0
5.	Credits: 4	6. Sem	ester: Both	7. Subject Area: PEC		

- 9. Pre-requisite: Nil
- **10. Objective:** The objective of this course is to introduce the fundamental techniques and algorithms used for acquiring, processing and extracting useful information from digital images.

11. Details of the Course

S.No.	Contents			
		Hours		
1.	Introduction: Signal processing overview; Image processing basics; Fundamental signals (1 D and 2 D): Classification of systems: Characteristics	4		
	of LTI/LSI systems. Introduction to the DIP areas and applications.			
2.	Digital Image Fundamentals: Human visual system and visual perception; Image sensing and acquisition Image file types; Pixel representation and spatial relationship	4		
3.	Image Digitization: Sampling and quantization. Image Transforms: 2- D DSFT and 2-D DFT, 2-D discrete cosine transform (DCT), 1-D and 2-D Karhonen Loeve (KL) or principal component analysis (PCA) and 1-D and 2-D discrete wavelet transforms and relation to filter banks.	8		
4.	Image Enhancement: Point and algebraic operations, edge detection and sharpening, filtering in the spatial and transformed domains. Rotation, interpolation, image filtering, spatial operators, morphological operators.	6		
5.	Image Segmentation: Thresholding; Edge based segmentation; Region growing; Watershed transform. Image Restoration: Degradation models, inverse and pseudo-inverse filtering, 2-D Wiener filtering and implementation	6		
6.	Image Compression and Encoding: Entropy-based schemes, Transform-based encoding, Predictive encoding and DPCM, Vector quantization, Huffman coding.	4		
7.	Feature Extraction and Segmentation: Contour and shape dependent feature extraction, textural features, region-based and feature-based segmentation.	5		
8.	Pattern Classification: Standard linear and Bayesian classifiers, supervised Vs unsupervised classification, classification performance index. Applications in satellite, sonar, radar and medical areas.	5		
	Total	42		

S.No.	Name of Authors/Book/ Publisher	Year of Publication/Reprint
1.	Gonzalez R. C. and Woods R. E., "Digital image processing," FourthEdition, Prentice Hall.	2017
2.	Lim J. S., "Two-dimensional signal and image processing," Prentice Hall.	1990
3.	Dudgeon D.E. and Merserau R. M., "Multidimensional digital signalprocessing," Prentice Hall Signal Processing Series.	1984
4.	Bose T., "Digital Signal and Image Processing", Wiley India.	2010
5.	Sonaka M., Hlavac V. and Boyle R., "Image Processing, Analysis and Machine Vision," Fourth edition, Cengage India Private Limited.	2017
6.	W.K. Pratt. "Digital Image Processing," Fourth Edition, John Wiley & Sons, New York.	2007

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

- 1. Subject Code: AID-578Course Title: Graphs Algorithms in Data Science
- **2. Contact Hours:** L: 3 T: 1 P: 0
- **3. Examination Duration (Hrs.): Theory:** 3 **Practical:** 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 46. Semester: Both7. Subject Area: PEC
- 8. Pre-requisite: Nil
- **9. Objective**: To acquaint the students with the knowledge of graph algorithms in ML and Data Science.

10. Details of the Course:

S.No.	Contents	
1	Fundamentals: graph models, Isomorphic graphs, Spanning tree, connectivity in graphs, Eulerian and Hamiltonian Graphs, matching, vertex colouring and domination, random graphs.	5
2	Graph Modelling with Neo4j: Graph Databases, directed vs undirected, weighted vs unweighted, cyclic vs acyclic, dense vs sparse, connected vs disconnected, graph traversal, Cypher Query Language, nodes and relationships, managing databases with Neo4j, creating, selecting a node, filtering, creating a relationship, selecting relationship, updating and deleting nodes and relationships, pattern matching and data retrieval, aggregation functions, importing data from CSV to JSON, Empowering business with pure Cypher, knowledge graphs, graph-based search, recommendation engines.	14
3	Graph Algorithms: The Graph Data Science Library and Path finding, Dijkstra's shortest path algorithm, A-star algorithm, k-shortest path, optimizing processes using graphs, travelling salesman problem, spanning tress, prims algorithm, minimum spanning tree in a Neo4j graph.	10
4	Spatial data: Node importance, representation spatial attributes, creating a geometry layer with Neo4j, spatial queries, visualization spatial data with Neo4j, Community detection and similarity measures.	8
5	Machine Learning on Graphs: Using graph-based features in machine Learning, predicting relationships, graph embedding from graphs to matrices, Applications of Neo4j in web applications.	5
	Total	42

S.No.	Name of Authors/Book/ Publisher	Year of Publication/Reprint
1	Jonathan Gross and Jay Yellen, Graph Theory and its	2018
	Applications, SecondEdition, CRC Press.	
2	Estelle Scifo, Hands-On Graph Analytics with Neo4j, Kindle Edition.	2020
3	Bondy J.A. and Murty U.S.R., Graph Theory I, Springer.	2013
4	Bela Bollobas, Random Graphs, Cambridge University Press.	2008
5	Douglas B. West —Graph Theory, Prentice Hall.	2014

NAME OF DEPARTMENT/CENTRE: Department of Computer Science and Engineering

- **1. Subject Code:** CSN-527 **Course Title**: Internet of Things
- 2. Contact Hours:L: 3T: 1P: 03. Examination Duration (Hrs.):Theory: 3Practical: 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 46. Semester: Both7. Subject Area: PEC
- 8. **Pre-requisite:** Knowledge of computer networks
- **9. Objective:** To impart the know-how of Internet of Things and their applications, architectures and protocols, building IoT applications/systems, securing the IoT systems, and their recent advances.

10. Details of the Course

S.No.	Contents	Contact hours
1.	Basic concepts revisited: Introduction to sensing & actuating, Basic networking, Wireless networks, Wireless sensor networks (WSN), Communication protocols, and other enabling technologies, IoT standards, Data storage & management issues and approaches, Cloud computing, Key challenges, research, and future directions of IoT, and security & privacy issues.	7
2.	Embedded Systems: Hardware and software of IoT, Microcontrollers, Understanding and programming Arduino, Raspberry Pi, NodeMCU, Lora, etc. Integrating microcontrollers with sensors and actuators, Building the IoT applications with any microcontroller.	6
3.	IoT Architectures and Protocols: Layers of communication, Architectures: State-of-the-art, IoT architecture reference models, Different views of IoT architectures and frameworks design, Protocols: Application protocols, Service discovery protocols, Infrastructure protocols, and other protocols. Understanding various types of protocols like HTTP, MQTT, CoAP, AMQP, 6LoWPAN, etc. Cross-layer implementations, and Data dissemination.	9
4.	Support Technologies for IoT: Big Data, Data Analytics, Artificial Intelligence, Mobile, Cloud, Software defined networks, 5G, and Fog/Edge computing. IoT integration with recent technologies. State-of-the-art. Design goals, challenges, and components.	8
5.	Cyber Physical Systems: Industry 4.0, Society 5.0, Design & use cases, Development, and implementation insights some examples like smart cities, smart homes, smart grids, smart agriculture, smart healthcare, smart transportation, smart manufacturing, and other smart systems. State-of-the-art. Conceptualizing the new IoT-based smart systems using a case study.	6
6.	IoT Security & Privacy: –, IoT Security and Privacy issues and challenges, Risks involved with IoT infrastructures, Trust in IoT platforms and other integrating technologies, Data aggregation, storage, retrieval, and other management issues including fault tolerance, interoperability, security, and privacy, Cyber-physical-systems and their security and privacy, Mitigation approaches.	6
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/ Reprint
1.	Edited by: Buyya, Rajkumar, and Amir Vahid Dastjerdi, Internet	2016
	of Things: Principles and paradigms. Elsevier/Morgan Kaufmann	
2.	Bahga, Arshdeep; Madisetti, Vijay, Internet of Things (A Hands-	2014
	on-Approach), AbeBooks.com	
3.	Sohraby, Kazem, Daniel Minoli, and Taieb Znati. Wireless sensor	2007
	networks: technology, protocols, and applications. John Wiley &	
	Sons	
4.	Marinescu, Dan C., Cloud computing: theory and practice.	2017
	Elsevier/ Morgan Kaufmann	

NAME OF DEPARTMENT/CENTRE: Department of Computer Science and Engineering

- **1. Subject Code**: AID-579 **Course Title**: Leveraging Data Science for Finance
- **2.** Contact Hours: L: 3 T: 0 P: 0
- **3. Examination Duration (Hrs.):** Theory: 3 Practical: 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 3 6. Semester: Both 7. Subject Area: PEC
- 8. Pre-requisite: Nil
- **9. Objective:** The objective of this course is to understand and apply the knowledge of data science related applications in the domain of finance.

10. Details of the Course:

S.No.	.No. Contents	
		Hours
1	Data Science basics: Preparation, organizing, and visualization of financial market	4
	data and examination of basic properties of security prices	
2	Quantiative models of risk-return framework in financial markets	4
3	Linear and non-linear price dynamics and modelling of security prices	4
4	4 Stock market prediction modelling, portfolio optimization, and wealth market	
	maximization	
5	5 Application of latent factor and commonality models in financial markets	
6	Modelling of financial market volatility using Conditional Heteroscedastic Models	6
7	7 Introduction to Crisis/Non-crisis models (Non-linearity, extreme-value modelling,	
	Markov regime-switching models, Quantile regression, Contagion models)	
8	Introduction to data modelling for high-frequency algorithmic trading	5
9	Use cases for application of data science in Finance: Investment Management,	4
	Sharpe ratio analysis, Capital Asset Pricing Model, etc. (using R programming)	
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/Reprint
1	Chris Brooks "Introductory Econometrics," Fourth Edition, Cambridge	2010
	University Press	2017
2	Ruey S. Tsay "Analysis of Time-series data," Third Edition, Wiley	2014
3	John Fox and Sanford Weisberg "An R Companion to Applied	2018
	Regression," Third Edition, SAGE	
4	Yves Croissant and Giovanni Millo "Panel Data Econometrics with R,"	2018
	First Edition, Wiley	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

- 1. Subject Code: AID-580 Course Title: Multi-Objective and Multi-Criteria Decision Making
- **2. Contact Hours:** L: 3 T: 1 P: 0
- **3. Examination Duration (Hrs.): Theory:** 3 **Practical:** 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 46. Semester: Both7. Subject Area: PEC
- 8. Pre-requisite: Nil
- **9. Objective:** To introduce the concept of multi-objective and multi criteria decision making systems to the students.

10. Details of the Course:

S.No.	Contents	Contact Hours		
1	1 Introduction: Review of decision making process in optimization and operations research models; overview of machine learning algorithms; ranking methods.			
2	Multi Objective Optimization (MOO): Introduction to multi objective	10		
	optimization, classical and recent methods for multi objective optimization			
	like genetic algorithms and particle swarm optimization.			
3	Multi Criteria Decision Making (MCDM): Introduction to MCDM methods; group decision making, weighing methods and ranking methods.	10		
4	Data Manipulation: Data wrangling and data management for large sized multi objective and multi criteria problems.	10		
5	Implementation: Implementation of the models developed in 2, 3 and 4 in	4		
	Python			
	Total	42		

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/Reprint
1.	G.H. Tzeng, J.J. Huang, "Multiple Attribute Decision	2011
	Making:Methods and Applications", CRC Press	
2.	M. Köksalan. J. Wallenius, S. Zionts, "Multiple Criteria	2011
	Decision Making. From Early History to the 21st Century",	
	World Scientific	
3.	J. Branke, K. Deb, K. Miettinen, R. Slowinski (Eds.),	2008
	"Multiobjective Optimization: Interactive and Evolutionary	
	Approaches", Springer-Verlag, Berlin, Heidelberg	
4.	A. Ishizaka, P. Nemery, "Multicriteria Decision Aid: Methods	2013
	and software", Wiley.	
5.	K Deb, "Multi-Objective Optimization Using Evolutionary	2011
	Algorithms", Wiley.	
6.	Michael Carter, Camille C. Price and Ghaith Rabadi	2018
	"Operations Research, A Practical Introduction", CRC Press.	

NAME OF DEPARTMENT/CENTRE: Department of Mathematics

- 1. Subject Code: MAN-634 Course Title: Parallel Computing
- **2. Contact Hours:** L: 3 T: 0 P: 0
- **3. Examination Duration (Hrs.): Theory:** 3 **Practical:** 0
- 4. Relative weightage: CWS: 20-35 PRS: 0 MTE: 20-30 ETE: 40-50 PRE:0
- 5. Credits: 36. Semester: Both7. Subject Area: PEC
- 8. Pre-requisite: Nil
- 9. Objective: To provide knowledge about parallel computing.
- **10. Details of the Course:**

S.No.	Contents	Contact Hours
1	Introduction, history, temporal parallelism, data parallelism, combined	
	temporal and data parallelism, data parallelism with dynamic and quasi-	
	dynamic assignment, specialist data parallelism, coarse-grained specialized	
	temporal parallelism, agenda parallelism, task dependencies and task graphs.	
2	Structures of parallel computers: classification of parallel computers based	8
	on data / instruction flow, coupling, mode of accessing memory, grain size,	
	vector supercomputers, systolic processors.	
3	Shared memory parallel computers based on shared bus& intercommunication	5
	networks, direct and indirect networks.	
4	Message Passing Systems, MPI Programming, point-to-point communications,	6
	collective communications	
5	CUDA Programming, host, device, threads, blocks, indexing, synchronization,	6
	performance optimization.	
6	Performance evaluation, parallel balance point, concurrency, scalability,	5
	speedup, Amdahl's law, Gustafson's law, Sun and Ni's law.	
7	Parallel algorithms, matrix multiplication, system of linear equations,	5
	sorting, discrete Fourier transforms, numerical integration.	
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of Publication/Reprint
		-
1	Aki, Selim G.: "The Design and Analysis of Parallel Algorithms", Prentice Hall, Englewood Cliffs, New Jersey.	1989
2	Krik, David B. and Hwu, W.W.: "Programming Massively	2010
	Parallel Processors - A Hands on Approach: Applications of	
	GPU ComputingSeries", Elsevier Inc.	
3	Pacheco, Peter S.: "Parallel Programming with MPI", Morgan	1997
	Kaufmann Publishers, Inc., California.	
4	Quinn, M. J.: "Parallel Computing: Theory and Practice", Tata	1994
	McGraw Hill.	
5	Rajaraman, V and Murthy, C. Siva Ram: "Parallel Computers	2000
	Architecture and Programming", Prentice Hall of India.	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

1.	Subject Code: AID-58	31		Course	e Title: Pattern l	Recognition
2.	Contact Hours:	L: 3	T: 1		P: 0	
3.	Examination Duration	n (Hrs.):	Theory: 3	Prac	t ical: 0	
4.	Relative Weightage:	CWS: 20-35	PRS: 0	MTE: 20-30	ETE: 40-50	PRE: 0
5.	Credits: 4	6. Semes	ster: Both	7. Subject Area: PEC		PEC
~						

- 8. Pre-requisite: Nil
- 9. Objective: To introduce various pattern recognition algorithms.

10. Details of the Course

S.No.	Contents	
		hours
1.	Introduction to Pattern Recognition and Bayesian Theory: Pattern	8
	recognition systems, The design cycle, Modeling using continuous and discrete	
	features, Discriminant functions, The Gaussian density, Error estimation, Some	
	basic examples	
2.	Parametric Models: Maximum-likelihood estimation, Bayesian estimation,	6
	Expectation-Maximization and mixture density estimation, Hidden Markov	
	Models, Bayesian Belief Networks	
3.	Non-parametric Methods and Feature Reduction: Density estimation,	8
	Parzen windows estimation, Nearest neighbor estimation, Curse of	
	dimensionality, Principal Component Analysis, Linear Discriminant Analysis,	
	Feature selection	
4.	Non-Bayesian Classifiers and Clustering: K-nearest neighbor classifier,	8
	Linear discriminant functions, Support vector machines, Neural networks,	
	Decision trees, Random Forests, Criterion functions for clustering, k-means	
	clustering, Hierarchical clustering, Graph-theoretic clustering, Cluster validity	
5.	Algorithm-Independent Learning Issues: No Free Lunch Theorem,	6
	Resampling for classifier design, Comparing classifiers, Combining classifiers	
6.	Structural and Syntactic Pattern Recognition: Recognition with strings,	6
	Grammatical methods, Graph-theoretic methods	
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of Publication/Reprint
1	P. O. Dude, P. F. Hart, D. G. Stork, Pattern Classification 2nd	2000
1.	R. O. Duda, P. E. Hall, D. G. Stork, Pattern Classification, 2	2000
	edition, John Wiley & Sons, Inc	
2.	C. M. Bishop, Neural Networks for Pattern Recognition,	1995
	Oxford University Press	
3.	K. Fukunaga, Introduction to Statistical Pattern Recognition,	1990
	Academic Press	
4.	R. Schalkoff, Pattern Recognition: Statistical, Structural and	1992
	Neural Approaches, John Wiley & Sons, Inc.	
5.	A. K. Jain, R. C. Dubes, Algorithms for Clustering Data, Prentice	1988
	Hall	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

- 1. Subject Code: AID-582 Course Title: Recommender Systems
- **2.** Contact Hours: L: 3 T: 0 P: 2
- **3. Examination Duration (Hrs.): Theory:** 3 **Practical:** 0
- **4. Relative Weightage: CWS:** 10-25 **PRS:** 25 **MTE:** 15-25 **ETE:** 30-40 **PRE:** 0
- 5. Credits: 46. Semester: Both7. Subject Area: PEC
- 8. Pre-requisite: Nil
- 9. Objective: The objective of this course is to learn and understand the algorithms, theories, and designs of recommender systems with relevant use cases.

10. Details of the Course:

S.No.	Contents	Contact	
		Hours	
1	Basic concepts for recommender systems, Detailed taxonomy of recommender		
	systems, Evaluation of recommender systems		
2	Content-based filtering algorithms, Collaborative filtering algorithms	6	
3	Neighborhood-based collaborative filtering algorithms (Memory-Based	6	
	Algorithms)		
4	Model-Based Collaborative Filtering Algorithms and Dimensionality Reduction	8	
5	Ensemble-Based and Hybrid Recommender Systems	6	
6	Advanced Topics in Recommendation Systems: The Cold Start, Context-aware	12	
recommender systems, time-sensitive, location-sensitive, social, and multi-criteria			
	Total	42	

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/Reprint
1	Charu Aggarwal "Recommender Systems: The Textbook," First	2016
		2015
2	Francesco Ricci, Lior Rokach, and Bracha Shapira "Recommender	2015
	SystemsHandbook," First Edition, Springer	
3	Rounak Banik "Hands-On Recommendation Systems with Python,"	2018
	First Edition, Packt Publishing	
4	Kim Falk "Practical Recommender Systems," First Edition,	2019
	Manning Publications	
5	Deepak Agarwal and Bee-Chung Chen "Statistical Methods for	2016
	RecommenderSystems," First Edition, Cambridge University Press	

NAME OF DEPARTMENT/CENTRE: Department of Computer Science and Engineering

- 1. Subject Code: CSN-519 Course Title: Social Network Analysis
- **2. Contact Hours:** L: 3 T: 1 P: 0
- **3. Examination Duration (Hrs.):** Theory: 3 Practical: 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 46. Semester: Spring7. Subject Area: PEC
- 8. Pre-requisite: Knowledge of computer networks
- 9. Objective: To introduce the basic notions used for social network analysis.

10. Details of the Course:

S.No.	Contents	Contact hours
1.	Social Network Analysis: Preliminaries and definitions, Erdos Number Project, Centrality measures, Balance and Homophily.	4
2.	Random graph models: Random graphs and alternative models, Models of network growth, Navigation in social Networks	4
3.	Network topology and diffusion, Contagion in Networks, Complex contagion, Percolation and information, Epidemics and information cascades	4
4.	Cohesive subgroups, Multidimensional Scaling, Structural equivalence, Roles and positions, Ego networks, Weak ties, Structural holes	6
5.	Small world experiments, Small world models, Origins of small world, Heavy tails, Small Diameter, Clustering of connectivity	6
6.	The Erdos Renyi Model, Clustering Models, Preferential Attachment	6
7.	Navigation in Networks Revisited, Important vertices and page rank algorithm, Towards rational dynamics in networks, Basics of game theory	6
8.	Coloring and consensus, biased voting, network formation games, network structure and equilibrium, behavioral experiments, Spatial and agent-based models	6
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication / Reprint
1.	Wasserman, Stanley, and Joseph Galaskiewicz. Advances in social	1994
	network analysis: Research in the social and behavioral sciences.	
	Sage	
2.	Knoke, David, and Song Yang. Social network analysis. Sage	2019
	Publications.	
3.	Carrington, Peter J., John Scott, and Stanley Wasserman, eds.	2005
	Models and methods in social network analysis. Vol. 28.	
	Cambridge university press.	
4.	Liu, Bing. "Social network analysis." In Web data mining, pp. 269-	2011
	309. Springer, Berlin, Heidelberg.	

NAME OF DEPARTMENT/CENTRE: Centre for Artificial Intelligence and Data Science

- 1. Subject Code: AID-583 Course Title: Data-driven Analytics for Smart Transportation Systems
- **2. Contact Hours:** L: 3 T: 1 P: 0
- **3. Examination Duration (Hrs.):** Theory: 3 Practical: 0
- **4. Relative Weightage: CWS:** 20-35 **PRS:** 0 **MTE:** 20-30 **ETE:** 40-50 **PRE:** 0
- 5. Credits: 46. Semester: Both7. Subject Area: PEC
- 8. Pre-requisite: Nil
- **9. Objective:** To familiarize with the applications of data science in traffic and transportation engineering and to demonstrate the applications of the data science in smart transportation planning

10. Details of the Course:

S.No.	Contents	
1	Data Science in Transportation	1001S
1.	Overview and Practical Applications: Transportation Data Sources: Data	•
	Collection: Data Preparation and Visualization.	
2.	Sensing and Data Mining for Smart Transportation Systems	8
	Intelligent Transportation Systems, Incident Management Program, Efficient	
	Emergency Vehicle Movement (Pre-Emption), Crash Detection, Reporting, and	
	Clearance; Traffic Surveillance, Identification of Hotspots, Violation Detection;	
	Road Network Asset Management, Inventory of Potholes, other Deficiencies;	
	Adaptive Traffic Signal.	
3.	Data Analytics in Urban Transportation Planning	10
	Basics of Urban Transportation Planning, Data Collection and Advanced Data	
	Sources, Household Surveys, Demand Modeling using WiFi/ Bluetooth/ Call Data	
	Record, Data Extraction and Analysis using APIs, Trip Distribution Modelling	
	Approaches, Route Choice Models, Choice Set Generation Methods, Genetic	
	Algorithms, Transportation Planning Example using Data-Driven Modeling and	
	Simulation.	
4.	Urban Mass Transit System	6
	Basics of Urban Mass Transit System, Static and Dynamic GTFS, Real-Time	
	Transit, Travel Time Variability, Transit Reliability, Transit Planning using Smart-	
	Card Data, Real-Time Accessibility Analysis.	
5.	Applications in Environmental Impact of Transport System	6
	IOT based Air pollution, Real-Time Air Pollution Monitoring and Data Analysis,	
	Placement of Mobile Sensors, Pollution Prediction using ML, Noise Data, Analysis	
	of Key Parameters, Development of Policy Framework.	
6.	Crash Data Analytics	8
	Crash Data, Data Preparation, Model Estimation, Real-Time Data-Driven	
	Analysis; Emergency Vehicle Data, Crash Prone Stretches, Ambulance	
	Deployment; Near-misses/Traffic Conflict Data, Surrogate Approach, Proactive	
	Assessment and Safety Interventions.	
	Total	42

S.No.	Name of Authors/Book/Publisher	Year of
		Publication/ Reprint
1.	Fumitaka Kurauchi, Jan-Dirk Schmöcker "Public transport planning	2021
	with smart card data" CRC Press	
2.	Juan de Dios Ortúzar, Luis G. Willumsen "Modelling Transport",	2011
	Wiley	
3.	Vukan R. Vuchic "Urban Transit: Operations, Planning, and	2005
	Economics" Wiley	
4.	Constantinos Antoniou, Loukas Dimitriou, Francisco Pereira	2018
	"Mobility Patterns, Big Data and Transport Analytics" Elsevier	
5.	Sara Moridpour, Alireza Toran Pour, Tayebeh Saghapour "Big Data	2019
	Analytics in Traffic and Transportation Engineering: Emerging	
	Research and Opportunities" IGI Global	
6.	Khaled R. Ahmed, Aboul-Ella Hassanien "Deep Learning and Big	2021
	Data for Intelligent Transportation" Springer	
7.	Davy Janssens, Ansar-Ul-Haque Yasar and Luk Knapen "Data	2013
	Science and Simulation in Transportation Research" IGI Global	