

GANPAT UNIVERSITY									
FACULTY OF ENGINEERING & TECHNOLOGY									
Programme		Master of Technology				Branch/Spec.		Computer Engineering (Artificial Intelligence)	
Semester		I				Version		1.0.0.0	
Effective from Academic Year			2025-26			Effective for the Batch admitted in			July 2025
Course Code		3CEAI101		Course Name		Foundations of AI			
Teaching Scheme						Examination Scheme (Marks)			
(Per week)		Lecture (DT)		Practical (Lab.)		Total			
		L	TU	P	TW			CE	SEE
Credit		3	-	1	-	4	Theory	50	50
Hours		3	-	2	-	5	Practical	30	20
Pre-requisites									
Basic engineering mathematics and coding									
Course Outcomes									
On successful completion of the course, the students will be able to:									
CO1	Understand and explain fundamental data structures and algorithmic techniques such as trees, heaps, hash tables, graph algorithms, and sorting/searching methods.								
CO2	Apply algorithmic design paradigms like divide-and-conquer, dynamic programming, and greedy algorithms to solve computational problems efficiently.								
CO3	Analyze and evaluate linear algebra and calculus-based techniques (e.g., matrix decomposition, gradient descent, eigen analysis) for optimization and mathematical modeling in computing.								
CO4	Construct probabilistic models using distributions, Bayes theorem, and statistical inference techniques for solving data-driven problems.								
Theory Syllabus									
Unit	Content								Hrs.
1	<b>Introduction to Data Structures and Algorithms</b> Basic data structures (linked list, stack, queue, trees, heap), Definition and elements of algorithms, characterizing running times, asymptotic running times								3
2	<b>Searching and Sorting</b> Binary Search Tree, Priority Queue, Sorting algorithms, Hash tables and hashing techniques								6
3	<b>Graph Algorithms</b> Graphs, Depth-first and Breadth-first search, Minimum Spanning Tree algorithms, Shortest-path algorithms								6
4	<b>Algorithm Design Paradigms</b> Divide-and-Conquer, Dynamic Programming, Greedy Algorithms with examples								6
5	<b>Matrix Algebra</b> Systems of linear equations, vector spaces, subspaces, linear dependence, independence, basis, linear transformations, rank, Inner product, norms of a vector, orthogonality, orthonormal basis, cosine, similarity, orthogonal matrices, eigenvalues and eigenvectors, Spectral, decomposition for real symmetric matrices, SVD								8
6	<b>Calculus and Optimization</b> Differentiation of functions of single and several variables, gradient and Hessian, Taylor's formula, Gradient of vector valued functions, higher order derivatives, linearization and multivariate Taylor's formula, Optimization of functions of several variables, characterization of stationary points, Constrained optimization, Lagrange multiplier approach Gradient free optimization methods for non-smooth optimization problems								8
7	<b>Probability</b> Axiomatic definition of probability, probability rules, Bayes Theorem, Random variables, pmfs and pdfs, CDF, Expectations, moments, Law of large numbers and central limit theorem, Special distributions, comparing probability distributions								8
Practical Content									
Practical, assignments and tutorials are based on above syllabus.									
Text Books									

1	Thomas H Cormen, Charles E Lieserson, Ronald L Rivest and Clifford Stein, Introduction to Algorithms, 4th Edition, MIT Press, 2022.
2	Jon Kleinberg and Eva Tardos, Algorithm Design, 1st Edition, Pearson, 2005.
3	Mathematics of Machine Learning, M. P. Deisenroth, A. A. Faisal, C. S. Ong, Cambridge University Press, 2020.
4	Linear Algebra and Optimization for Machine Learning, Charu C. Aggarwal, Springer, 2020
<b>Reference Books</b>	
1	1) Sanjoy Dasgupta, Christos H. Papadimitriou and Umesh V. Vazirani, Algorithms, Tata McGraw-Hill, 2008.
2	Michael T Goodrich and Roberto Tamassia, Algorithm Design: Foundations, Analysis, and Internet Examples, Second Edition, Wiley, 2006.
3	Optimization for Data Analysis, S. Wright and B. Recht, Cambridge University Press, 2022.
4	Introduction to Applied Linear Algebra: Vectors, Matrices and Least Squares, S. Boyd and L. Vandenberghe, Cambridge University Press, 2018.
<b>ICT/MOOCs Reference</b>	
1	<a href="https://onlinecourses.nptel.ac.in/noc25_cs136/preview">https://onlinecourses.nptel.ac.in/noc25_cs136/preview</a>

	P O 1	P O 2	P O 3	P O 4	P O 5	P O 6	P O 7	P O 8	P O 9	P O 10	P O 11	P S O 1	P S O 2	P S O 3
CO1	3	2	1	1	1	0	1	0	1	0	0	2	2	2
CO2	3	2	1	1	1	0	1	0	1	0	0	2	2	2
CO3	3	3	2	3	1	1	3	0	2	0	1	3	2	3
CO4	3	3	2	3	3	1	3	0	2	0	1	3	3	3