# Revised Syllabus for Two Years Program (Under the New Education Policy, NEP 2020) in

M.Sc. in Engineering Mathematics (2023-2024)



# DEPARTMENT OF MATHEMATICS INSTITUTE OF CHEMICAL TECHNOLOGY

(University Under Section-3 of UGC Act, 1956)

Elite Status and Center for Excellence

Government of Maharashtra

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### A. Preamble:

Due to the emergence of modern computing facilities, the applications of mathematics in all branches of engineering, medical sciences, and financial sectors, etc. have become extremely important. As a result, there has been an extraordinary demand for technically qualified professionals having sound mathematical skills. However, most of the courses available in the country are devoted to either pure mathematics or some combination of pure and applied mathematics with a bit of computer programming knowledge.

The master's program entitled "M.Sc. in Engineering Mathematics", offered by the Department of Mathematics, Institute of Chemical Technology, Mumbai, is designed to create professionals who are equipped with practical knowledge of Mathematics, Statistics and Computer Programming. This is a two-year programme consisting of four semesters. The programme was started as a five-year project under UGC Innovative Schemes in 2011. The first revision of the syllabus took place in the academic year 2017-2018. The second revision has been done during the academic year 2021-2022. This is the third revision which will be implemented from the academic year 2023 – 2024. The revision has been carried out as per the guidelines of the National Education Policy 2022.

The course is an optimal blend of mathematical theory and its applications. Subjects related to Applied Mathematics, Statistics and Machine Learning will train the students on the use of modern computational tools to solve real life problems which are relevant to industry and society. These applied courses are complemented by some foundation courses in pure mathematics. Students having gone through this course will have sound working knowledge with a strong mathematical base which is necessary to address computational and statistical challenges encountered in the different areas of science and technology. During the course, students also work on a yearlong project during the third and fourth semester under the supervision of faculty members of the department and most often the projects are carried out in collaboration with people from industry.

After completing this course students will have career opportunities both in industry and academia. Almost all the career paths open to graduates in Mathematics are also available to the students.

# B. Regulations Related to the Degree of Master of Science in Mathematics (M. Sc. in Engineering Mathematics) Degree Course

# • Intake

20 candidates shall be admitted every year. The distribution of seats shall be as per the Institute's norms.

### Admission

- a. Candidates who have taken the post-H.S.C. 3-year/4-year degree course of Bachelor of Science with 6 units of Mathematics at the third year of the course and any two of chemistry, physics, and statistics as the two other subjects at the first and second years of University of Mumbai or of any other recognized University; and passed the qualifying examination with at least 55% of the marks in aggregate or equivalent grade average. (50% for the backward class candidates only from Maharashtra State are eligible to apply).
- b. Candidates who have passed B.Sc. in Statistics or B.Tech./B.E. with at least four mathematics papers as part of the coursework from a UGC/AICTE recognized University/ Institute are also eligible for admission.
- c. The admissions will be done strictly based on merit; the marks obtained in entrance test conducted by ICT.
- d. The candidates who have cleared the qualifying examination in one sitting will be preferred.

## Course structure

- a. The course is a credit-based 4-semester (2-year) course.
- b. The course has an exit option after one year with a "Diploma" as per the guidelines of NEP 2020.

- c. There will be two semesters in a year:
  - i. Semester I and Semester III (July to December)
  - ii. Semester II and Semester IV (December to May)
  - iii. Each semester will consist of 15-16 weeks of instructions including seminars / projects/assignments.
- d. The On Job Training (OJT) will be at the end of second semester (during summer) for 8-10 weeks and carries 4 credits.
- e. At the end of each semester the candidates will be assessed as per the norms of the Institute.
- f. Semesters will be governed by academic calendar of the institute.
- g. The requirement of attendance of the students shall be as per the norms of the Institute.
- h. All the relevant academic regulations of the Institute shall be applicable to the course.
- i. Assessment of the students will be done as per the norms of the Institute.
- j. In case of any difficulty regarding any assessment component of the course, the Departmental Committee shall take appropriate decision, which will be considered final.

### k. Electives:

- i. Three elective courses will be offered during the programme and the list of electives will be made available to the students.
- ii. Open electives will also be offered which may be of two types: (a) students can take it from well-established MOOC courses with prior approval from the department (b) it may be proposed by a faculty with detailed syllabus and get prior approval from the department.

# 1. Research Project:

- i. At the end of the Second semester, the Head of Department in consultation with the Departmental Committee will assign topics for the Research Projects (4 credits) to the students and assign the supervisors.
- ii. The students will do the Research Projects (6 credits) in semester IV on the topics assigned under the supervision of the assigned faculty member.
- iii. The students shall submit the project report before the prescribed date which will be a date before the last date of the semester IV. The report shall be submitted with soft binding.
- iv. The project report will be examined by the supervisor along with one other internal/external referee to be appointed by the Departmental committee. The referees shall give marks to the report as per the norms.
- v. The students will make presentation on the work in front of the Project Evaluation Committee (PEC) appointed by the Departmental Committee, in open defence form. The PEC will give marks to the presentation.
- vi. The comments received from the referees as well as given by the PEC need to be incorporated in the final project report in consultation with the supervisor.

# **Programme Outcomes (POs) for M.Sc. Engineering Mathematics**

PO1	Fundamental knowledge of pure mathematics: Apply the fundamental concepts of pure mathematics to understand the concepts in Applied Mathematics, Statistics and Computational Mathematics and empowering students to engage in research and development in future into top industries and institutions.
PO2	Foundation of Applied Mathematics: Strong foundation of Applied Mathematics which is directly connected to solving real life problems in different domains by means of mathematical modelling and being able to apply them in solving complex problems relevant to the society and industry.
PO3	Foundation of Statistics and Data Science: Strong foundation of Mathematics and Statistics of Data science and good hold on various statistical methodologies including probability theory, estimation, and testing of hypothesis etc. and being able to apply them in solving real life problems.
PO4	Foundation of Machine Learning and AI: Understand and employ modern computational methods in Machine Learning including Deep Learning and Artificial Intelligence and use them effectively using free and proprietary advanced computational platforms for solving large scale problems arising from different research areas.
PO5	Research based Teaching Learning: An innovative teaching framework to engage students in both academic and industrial research and open up multiple future paths in different verticals including preparation to qualifying national level tests like NET/GATE etc and creation of future leaders in teaching.
PO6	Conduct investigations of complex problems: Use research-based knowledge in machine learning and artificial intelligence and research methods including design of experiments, analysis and interpretation of data to unfold complex problems from industry and academia and provide working solutions.
PO7	Problem analysis: Identify, formulate, review research literature, and analyze complex real life problems using mathematics, statistics, and computational methods.
PO8	Societal Applications of Mathematics: Apply reasoning informed by the existing knowledge pool to convert into a quantitative framework, collect relevant information and address various societal issues using modelling and statistical data analytics tools including deep learning and artificial intelligence.
PO9	Ethics: Apply ethical principles and commit to professional ethics and responsibilities and norms of the practice of mathematics, statistics, and data sciences in all verticals of industry and society.
PO10	Individual and teamwork: Function effectively as an individual, and as a member or leader in diverse teams, and in multidisciplinary settings.
PO11	Communication: Communicate effectively on complex industrial/natural problems and understand the functional requirements, identify the gaps and being able to provide solutions using modern tools and technologies offering advanced data sciences and machine learning techniques.
PO12	Life-long learning: Recognize the need for and have the preparation and ability to engage in independent and life-long learning, acquire appropriate skills in Mathematics and its application for the benefit of humankind.

# Programme Specific Outcomes (PSOs) for M.Sc. Engineering Mathematics

Practical Applications of Mathematics: Strong foundation of Applied Mathematics which is directly connected to solving real life problems in different domains by means of mathematical modelling and being able to apply them in solving complex problems.  Foundations of Statistics and Data Science: Gain a strong understanding on the Statistical foundations of Data Science and Machine Learning and apply them to effectively quantify the uncertainty in decision making in real life problems.  Foundations of Scientific Programming: Strong foundations on mathematical and probabilistic computations using free and open-source software and develop algorithmic thinking to address computational challenges.  Real World applications of Machine Learning and AI: Communicate effectively AI concepts and methodologies and gain proficiency in applying them in addressing real world problems coming from various domains such as healthcare, finance, environment and climate related applications, etc.  Collaborative and Interdisciplinary Problem Solving: Function effectively as an individual, and as a		
connected to solving real life problems in different domains by means of mathematical modelling and being able to apply them in solving complex problems.  Foundations of Statistics and Data Science: Gain a strong understanding on the Statistical foundations of Data Science and Machine Learning and apply them to effectively quantify the uncertainty in decision making in real life problems.  Foundations of Scientific Programming: Strong foundations on mathematical and probabilistic computations using free and open-source software and develop algorithmic thinking to address computational challenges.  Real World applications of Machine Learning and AI: Communicate effectively AI concepts and methodologies and gain proficiency in applying them in addressing real world problems coming from various domains such as healthcare, finance, environment and climate related applications, etc.  Collaborative and Interdisciplinary Problem Solving: Function effectively as an individual, and as a member in large scale data science projects in interdisciplinary research involving both academia and industry.	PSO1	sciences and learn how to apply mathematical reasoning in a wide range of theoretical and applied
of Data Science and Machine Learning and apply them to effectively quantify the uncertainty in decision making in real life problems.  Foundations of Scientific Programming: Strong foundations on mathematical and probabilistic computations using free and open-source software and develop algorithmic thinking to address computational challenges.  Real World applications of Machine Learning and AI: Communicate effectively AI concepts and methodologies and gain proficiency in applying them in addressing real world problems coming from various domains such as healthcare, finance, environment and climate related applications, etc.  Collaborative and Interdisciplinary Problem Solving: Function effectively as an individual, and as a member in large scale data science projects in interdisciplinary research involving both academia and industry.	PSO2	connected to solving real life problems in different domains by means of mathematical modelling and
PSO4 computations using free and open-source software and develop algorithmic thinking to address computational challenges.  Real World applications of Machine Learning and AI: Communicate effectively AI concepts and methodologies and gain proficiency in applying them in addressing real world problems coming from various domains such as healthcare, finance, environment and climate related applications, etc.  Collaborative and Interdisciplinary Problem Solving: Function effectively as an individual, and as a member in large scale data science projects in interdisciplinary research involving both academia and industry.	PSO3	of Data Science and Machine Learning and apply them to effectively quantify the uncertainty in
methodologies and gain proficiency in applying them in addressing real world problems coming from various domains such as healthcare, finance, environment and climate related applications, etc.  Collaborative and Interdisciplinary Problem Solving: Function effectively as an individual, and as a member in large scale data science projects in interdisciplinary research involving both academia and industry.	PSO4	computations using free and open-source software and develop algorithmic thinking to address
PSO6 member in large scale data science projects in interdisciplinary research involving both academia and industry.	PSO5	methodologies and gain proficiency in applying them in addressing real world problems coming from
demic Collincial Ox	PSO6	member in large scale data science projects in interdisciplinary research involving both academia and
K. T.		ademic

# M.Sc. in Engineering Mathematics (Under NEP 2020) Institute of Chemical Technology, Mumbai

	Se	mester-I							
Subject Code	Subject Credits Hrs/Week Marks for various E								
			L	T	P	CA	MS	ES	Total
MAT 2201	Applied Linear Algebra	4	4	0	0	20	30	50	100
MAT 2230	Real and Complex Analysis	4	4	0	0	20	30	50	100
MAT 2326	Statistical Computing	4	4	0	0	20	30	50	100
HUT2102F	Research Methodology	4	4	0	0	20	30	50	100
	Elective-I	4	4	0	0	20	30	50	100
MAP 2523	Computational Mathematics Lab – I	2	0	0	4		50	50	100
	Total	22	22	0	4				600

	Sei	mester-II							
Subject Code	Subject Credits Hrs/Week Marks for various Exa								
			L	T	P	CA	MS	ES	Total
MAT 2235	Differential Equations	4	4	0	0	20	30	50	100
MAT 2231	Modern Algebra	4	4	0	0	20	30	50	100
MAT 2327	Machine Learning	4	4	0	0	20	30	50	100
	Elective-II	4	4	0	0	20	30	50	100
MAP 2524	Computational Mathematics Lab – II	2	0	0	4		50	50	100
MAP 2811	On Job Training (OJT)	4						100	100
	Total	22	18	0	4				600

# Exit option after the second semester with PG Diploma Degree

	Sen	nester III							
Subject Code	Subject Credits Hrs/Week Marks for variou								
			L	T	P	CA	MS	ES	Total
MAT 2229	Measure, Integration and Functional Analysis	4	4	0	0	20	30	50	100
MAT 2232	Optimization Techniques	4	4	0	0	20	30	50	100
MAT 2328	Deep Learning and Artificial Intelligence	4	4	0	0	20	30	50	100
	Elective – III	4	4	0	0	20	30	50	100
MAP 2704	Research Project (RP)	4			8			100	100
MAP 2525	Computational Mathematics Lab – III	2	0	0	4		50	50	100
	Total	22	18	0	12				600

	Semester – IV										
Subject Code	Subject	Credits	Hrs/	Weel	ζ.	Mar	ks for v	arious	Exams		
			L	T	P	CA	MS	ES	Total		
MAT 2233	Advanced Differential Equations	4	4	0	0	20	30	50	100		

MAT 2329	Advanced Statistical Computing	4	4	0	0	20	30	50	100
MAP 2705	Research Project (RP)	6	0	0	12			100	100
MAT 2234	Mathematical Modelling	4	4	0	0	20	30	50	100
	Elective – IV	4	4	0	0	20	30	50	100
	Total	22	16	0	12				500

In each semester, the department will offer electives from the following list of topics.

	List o	of Elective	es						
Subject Code	Subject	Credits	Hrs/	Weel	ζ	Mar	ks for v	arious	Exams
			L	T	P	CA	MS	ES	Total
MAT 2651	Graph Theory	4	4	0	0	20	30	50	100
MAT 2612	Combinatorics	4	4	0	0	20	30	50	100
MAT 2606	Financial Mathematics	4	4	0	0	20	30	50	100
MAT 2603	Number Theory	4	4	0	0	20	30	50	100
MAT 2605	Groups and Symmetries	4	4	0	0 <	20	30	50	100
MAT 2607	Matrix Computations	4	4	0	0	20	30	50	100
MAT 2621	Cryptography	4	4	0	0	20	30	50	100
MAT 2608	Topology	4	4	0	0	20	30	50	100
MAT 2609	Stochastic Process	4	4	0	0	20	30	50	100
MAT 2630	Coding Theory	4	4	0	0	20	30	50	100
MAT 2649	Advanced Modern Algebra	4	4	0	0	20	30	50	100
MAT 2622	Finite Element Method	4	4	0	0	20	30	50	100
MAT 2642	Integral Transforms	4	4	0	0	20	30	50	100
MAT 2627	Mathematical Biology	4	4	0	0	20	30	50	100
MAT 2628	Signal Processing	4	4	0	0	20	30	50	100
MAT 2629	Momentum, Heat and Mass Transfer	4	4	0	0	20	30	50	100
MAT 2650	Representation Theory	4	4	0	0	20	30	50	100
MAT 2610	Advanced Mathematical Finance	4	4	0	0	20	30	50	100
MAT 2625	Multivariate Analysis	4	4	0	0	20	30	50	100
MAT 2626	Design and Analysis of Experiments	4	4	0	0	20	30	50	100
MAT 2623	Operation Research	4	4	0	0	20	30	50	100
MAT 2644	Geometry of Curves and Surfaces	4	4	0	0	20	30	50	100
MAT 2645	Convex Optimization	4	4	0	0	20	30	50	100
MAT 2646	Time-Series Analysis	4	4	0	0	20	30	50	100
MAT 2611	Computational Fluid Dynamics	4	4	0	0	20	30	50	100
MAT 2647	Operator Theory	4	4	0	0	20	30	50	100
	*Open Elective	4	4	0	0	20	30	50	100

<sup>\*</sup>Open electives may be of two types: (i) students can take it from MOOC (Swayam/NPTEL etc.) course with prior approval from the department (ii) it may be proposed by a faculty with complete details of syllabus and get prior approval from the department.

SEMESTER I

Approve by Acadel.

	Common Codes MAT 2201	Course Title: Applied Linear Algebra		dits=	: 4
	Course Code: MAT 2201	Course Tiue: Applied Linear Algebra	L	T	P
	Semester: I	Total contact hours: 60	4	0	0
		List of Prerequisite Courses			
Basics	s of matrix algebra and determin	ant of square matrix, vector spaces			
	List of Co	urses where this course will be prerequisite			
It is a	foundation course which will be	prerequisite for all the course studied in this program			
	Description of relevance of	this course in the M.Sc. Engineering Mathematics Progra	am	0	
This i	s a course further built up on an	d in continuation with undergraduate level course on linear	algeb	ra. Tl	nis

This is a course further built up on and in continuation with undergraduate level course on linear algebra. This course reviews the major concepts of linear algebra and introduces advanced concepts with real life applications. Introduced concepts which will be used in almost all the later courses with special emphasis on Machine Learning and Deep Learning concepts.

	Course Contents (Topics and subtopics)	Hours
1	Review of Vector Spaces, Subspaces, Linear dependence and independence, Basis and	6
1	dimensions.	6
	Basic concepts in Linear Transformations; Use of elementary row operations to find	
2	coordinate of a vector, change of basis matrix, matrix of a linear transformations and	8
	subspaces associated with matrices.	
3	Inner Product Spaces, Orthogonal Bases, Gram-Schmidt Orthogonalization, QR	12
3	Factorization, Normed Linear Spaces.	12
4	Matrix Norm, condition numbers and applications.	4
	Eigenvalue and Eigenvectors, Diagonalization and its applications to ODE, Dynamical	
5	Systems and Markov Chains, Positive Definite Matrices and their applications,	10
	Computation of Numerical Eigenvalues.	
	Singular Value Decomposition, Matrix Properties via SVD, Projections, Least Squares	
6	Problems, Application of SVD to Image Processing, Principal Component Analysis	10
	(PCA).	
	Structure of Linear Maps: Adjoint operators, Normal, Unitary, and Self-Adjoint	
7	operators, Spectral theorem for normal operators, Jordan Canonical Forms and its	10
	applications.	
	List of Textbooks/ Reference books	
1	S. Kumaresan, Linear Algebra – A Geometric Approach, Prentice Hall India.	
2	David C Lay, Linear Algebra and its Applications, Addition-Wesley.	
3	Richard Bronson and Gabriel B. Costa, Matrix Methods, Academic Press.	
4	G. Strang, Linear Algebra and its Applications, Harcourt Brace Jovanish.	
5	Robert Beezer, A First Course in Linear Algebra, open textbook (http://linear.ups.edu/html/	fcla.html)
6	Carl D. Mayer, Matrix Analysis and Applied Linear Algebra, SIAM.	
7	G. C. Cullen, Linear Algebra with Applications, Addison Wesley.	
	Course Outcomes (students will be able to)	
CO1	Understand concepts in Linear Transformations and Inner Product spaces	
CO2	Understand basic concepts in Eigenvalues-Eigenvectors and Structure of Linear maps.	
CO3	Understand and work with various matrix factorization.	
CO4	Apply applied linear algebra concepts to solve real life problems.	
CO5	Apply concepts in eigenvalues-eigenvectors to solve real life problems.	

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	3	3	3	2	1	1	0	3	1	0	3

CO2	3	3	3	3	0	0	1	0	3	0	0	3
CO3	3	3	3	3	0	0	1	1	3	0	0	3
CO4	3	3	3	3	2	3	2	1	3	2	2	3
CO5	3	3	3	3	2	3	2	2	3	2	2	3

N	<b>Iapping of Cou</b>	rse Outcomes (	COs) with Progr	ramme Specific	<b>Outcomes (PS</b>	Os)
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6
CO1	3	0	0	0	0	0
CO2	3	0	0	0	0	0
CO3	3	0	0	0	0	0
CO4	1	3	1	1	2	0
CO5	1	3	1	1	2	0

	Commo Codo, MAT 2220	Course Titles Deel and Compley Analysis	Credits =					
	Course Code: MAT 2230	Course Title: Real and Complex Analysis	L	T	P			
	Semester: I	Total contact hours: 60	4	0	0			
		List of Prerequisite Courses						
Basic c	course in Calculus							
		ourses where this course will be prerequisite						
	_	Analysis (MAT 2229), Advanced Differential Equations	(Ma	ΑT	2233),			
Operate	or Theory (MAT 2647)							
		of this course in the M.Sc. Engineering Mathematics Progr						
		requisite for all the pure and applied mathematics topics inclu	ıdin	g sta	itistics			
in upco	oming semesters			**				
		ontents (Topics and subtopics)	-	Ho	urs			
1	-	ctions, uniform convergence and its relation to continuity,		1	0			
		n. Weierstrass approximation theorem.	-					
		es, Convergence of sequences of several variables, Limits derivatives, Differentiability of functions from $\mathbb{R}^n$ to $\mathbb{R}^m$ ,						
2				2	0			
	Higher order derivatives, Taylor's theorem and application, Local Maxima, Local Minima, Saddle points, Stationary points.							
	- / /	uchy's theorems, Cauchy's integral formula, Liouville's						
3	theorem.	deny 5 dicoronis, Cadeny 5 integral formata, Liouvine 5		2	0			
		, isolated singularities and residues, Classification of						
4	singularities, Residue theory	,		1	0			
		List of Textbooks / Reference books						
1	T. Apostol, Mathematical An	alysis, 2nd Edition, Narosa, 2002.						
2	W. Rudin, Principles of Math	nematical Analysis, 3rd Edition, McGraw-Hill						
3	Ajit Kumar and S. Kumaresa	n, A Basic Course in Real Analysis, CRC Press.						
4	S. Kumaresan, A Pathway to	Complex Analysis, Techno World Publications						
5	T. M. Apostol, Calculus Vol.	II, 2nd Ed., John Wiely& Sons.						
6	J. E. Marsden, A. Tromba,	and A. Weinstein, Basic Multivariable Calculus, Springer-Ve	rlag.					
7	Susane Jane Colly, Vector Ca	alculus, 4th Edition, Pearson.						
8	J. B. Conway, Functions of C	One Complex Variable, 2nd Edition, Narosa, New Delhi.	_					
9	_	lysis, Springer International Edition.						
	Cour	rse Outcomes (students will be able to)						
CO1		l uniform convergence of sequence and series of functions.						
CO2	TT. 1 1.41 C. 1:00	erentiability from R <sup>n</sup> to R <sup>m</sup> .						

CO3	Obtain Taylor series expansions of functions of several variables and compute maxima,	
COS	minima and saddle points.	
CO4	Understand analytic functions and apply Cauchy's theorem to compute complex	
C04	integrals.	
CO5	Classify singularities of a function.	

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	3	0	1	2	0	2	1	3	1	0	3
CO2	3	3	1	2	2	0	1	1	3	1	0	73
CO3	3	3	1	2	2	1	2	0	3	0	0	3
CO4	3	3	1	1	2	2	3	1	3	1	0	3
CO5	3	3	0	1	2	2	3	1	3	0	0	3

	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6
CO1	3	0	0	0	0	0
CO2	3	0	0	0	0	0
CO3	3	0	0	0	0	0
CO4	3	0	0	0	0	0
CO5	3	0	0	0	0	0
			ic Coti	<b>,</b>		

	Course Code: MAT 2326 Course Title: Statistical Computing	C	redit	s = 4				
	Course Code: MAT 2326 Course Title: Statistical Computing	L	T	P				
	Semester: I Total contact hours: 60	4	0	0				
	List of Prerequisite Courses							
Basic o	course on Calculus							
	List of Courses where this course will be prerequisite							
Machin	ne Learning (MAT 2327), Advanced Statistical Computing (MAT 2329), Deep Learning	and	l Art	ificial				
Intellig	gence (MAT 2328), Stochastic Process (MAT 2609), Computational Mathematics Lab – II (N	ИAF	252	4)				
	Description of relevance of this course in the M.Sc. Engineering Mathematics Progr	am						
This co	ourse is a foundation course covering major concepts of Probability and Estimation Theo	ry.	Intro	duced				
concep	ts which will be used in all Machine Learning and Deep Learning courses.							
7	Course Contents (Topics and subtopics)		Hot	ırs				
1	Introduction to Probability: Random experiment, Probability space, Conditional							
1	Probability and Independence, Bayes Theorem 6							
	Random Variables and Their Probability Distributions: Random variables and their							
2	distributions. Discrete and Continuous random variables. Functions of random variables							

and their distribution, Common discrete distributions, Common continuous distributions Distribution of Functions of random variables (emphasis on transformation formula).

Moments and Generating Functions: Moments of distribution function, generating

functions (moment generating function, probability generating function, characteristic

function, cumulant generating function, factorial moment generating functions) and their

10

4

2

3

applications, Moment Inequalities.  Multiple Random Variables and Sampling distributions: Joint distribution, Independence, functions of several random variables, Covariance, Correlation and joint moments, Conditional Expectation. Concept of Random sampling, Sample characteristics and their distribution, Chi-Square, t-, and F-Distributions: Exact Sampling Distributions; Sampling from Normal distribution, Order Statistics, and their distributions  Limit theorems: Convergence concepts, Weak Law of Large Numbers and Strong Law of Large numbers, Central Limit Theorem  Elements of Estimation theory: Parametric Point estimation: Finding estimators using method of moments, maximum likelihood. Properties of estimators: Sufficiency, factorization theorem, Rao-Blackwell theorem. Unbiased estimates and uniformly minimum variance unbiased estimators. Fisher Information and Cramer-Rao inequality, comparing estimators based on risk function.  Elements of Hypothesis testing: Likelihood Ratio tests, Wald tests, Error probabilities and the power function, most powerful tests.  Tests related to normal distribution: Sampling from normal distribution and test for mean, tests on variance, tests on several means, and tests on several variances with practical problems and applications.  Interval Estimation: Inversion of test statistics, Size and coverage probability,	6 6 6
Flements of Estimation theory: Parametric Point estimation: Finding estimators using method of moments, maximum likelihood. Properties of estimators: Sufficiency, factorization theorem, Rao-Blackwell theorem. Unbiased estimates and uniformly minimum variance unbiased estimators. Fisher Information and Cramer-Rao inequality, comparing estimators based on risk function.  Flements of Hypothesis testing: Likelihood Ratio tests, Wald tests, Error probabilities and the power function, most powerful tests.  Tests related to normal distribution: Sampling from normal distribution and test for mean, tests on variance, tests on several means, and tests on several variances with practical problems and applications.  Interval Estimation: Inversion of test statistics, Size and coverage probability,	6
method of moments, maximum likelihood. Properties of estimators: Sufficiency, factorization theorem, Rao-Blackwell theorem. Unbiased estimates and uniformly minimum variance unbiased estimators. Fisher Information and Cramer-Rao inequality, comparing estimators based on risk function.  Flements of Hypothesis testing: Likelihood Ratio tests, Wald tests, Error probabilities and the power function, most powerful tests.  Tests related to normal distribution: Sampling from normal distribution and test for mean, tests on variance, tests on several means, and tests on several variances with practical problems and applications.  Interval Estimation: Inversion of test statistics, Size and coverage probability,	6
and the power function, most powerful tests.  Tests related to normal distribution: Sampling from normal distribution and test for mean, tests on variance, tests on several means, and tests on several variances with practical problems and applications.  Interval Estimation: Inversion of test statistics, Size and coverage probability,	
8 mean, tests on variance, tests on several means, and tests on several variances with practical problems and applications.  Interval Estimation: Inversion of test statistics, Size and coverage probability,	4
	4
Connection to Testing of hypothesis	4
Software component for module 8 and 9 will be covered in Research Methodology in Mathematical Sciences (HUT2012F)	
List of Textbooks / Reference Books	
P.G. Hoel, S.C. Port and C.J. Stone, Introduction to Probability, Universal Book Stall, New De	elhi.
2 K. Md. Ehsanes Saleh and V. K. Rohatgi. An Introduction to Probability and Statistics. Wiley.	
3 G. Casella and R. L. Berger. Statistical Inference. Duxbury Press.	
4 W. W. Hines, D. C. Montgomery, Probability and Statistics in Engineering. John Wiley.	
5 V. Robert Hogg, T. Allen Craig. Introduction to Mathematical Statistics, McMillan Publication	
6 Vijay K. Rohatgi and A. K. Md. Ehsanes Saleh, An Introduction to Probability and Stati Wiley & Sons, Inc.	stics, John
A. M. Mood, F. A. Graybill and D. C. Boes, Introduction to The Theory of Statistics, Third E Graw Hill Education.	dition, Mc
8 A. M. Gun, M. K. Gupta, B. Dasgupta, An Outline of Statistical Theory, Volume Two, World	Press.
9 L. Wasserman, All of Statistics: A Concise Course in Statistical Inference, Springer	
Course Outcomes (students will be able to)	
CO1 Compute probability of events for basic combinatorial problems	
CO2 Compute moments and distributions of random variables and functions of random variables	
CO3 Understand various convergence concepts and apply them to investigate large samples properties of estimators	
CO4 Estimate parameters of a population distribution using maximum likelihood and method of moments	
CO5 Understand different types of errors in testing of hypothesis and plot power functions.	
CO6 Apply basic testing procedure to solve data analysis problems	
CO7 Compute interval estimators for population parameters and apply it to solve real life problems.	

Mapping of Course Outcomes (COs) with Programme Outcomes (POs)												
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	1	2	3	3	2	2	3	2	3	1	1	3

CO2	1	2	3	3	1	2	3	2	3	1	0	3
CO3	1	2	3	3	1	2	3	2	3	1	0	3
CO4	1	1	3	3	1	2	3	1	3	1	0	3
CO5	1	3	3	3	2	3	3	2	3	1	0	3
CO6	1	2	3	3	1	2	3	3	3	2	2	3
CO7	1	2	3	3	1	3	3	1	3	2	2	3

M	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)								
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6			
CO1	0	0	3	0	1	42			
CO2	0	0	3	0	1	1			
CO3	0	0	3	0	1	1			
CO4	0	0	3	0	1	1			
CO5	0	0	3	0	1	1			
CO6	0	0	3	0	1	1			
CO7	0	0	3	0	61°	1			

Course Title: Research Methodology in Mathematical

Credits = 4

	Course Code: HUT2102F	Sciences	L	T	P				
	Semester: I	Total contact hours: 60	4	0	0				
		List of Prerequisite Courses							
NIL									
		Courses where this course will be prerequisite							
On Job	<u> </u>	Research Projects (MAP 2704, 2705)							
		e of this course in the Ph.D. in Mathematics Program at IO							
It is a Science		ng out research works at M.Sc. and Ph.D. programmes in	Matl	hema	tical				
	Course C	Contents (Topics and subtopics)	I	Hour	S				
		thodology, definition and characteristic of research, different							
1	types of research, Literature survey and formulation of research problem, Developing								
	objectives, Research designs, Data collection								
		software such as Python, R, SAGEMATH, Mathematica,							
2	Matlab.	P. Date to an Date level d'en late l'annual et le		1.5					
2		R: Data types, Data description, data import and export, a exploration and summary statistics, Histograms, boxplot,		15					
	A	cobability plot, quantile-quantile plot							
	1	Testing of Hypothesis: Discrete and continuous probability							
3		butions, basic testing procedures for real data analysis using		20					
	R/Python	buttons, busic testing procedures for real data untaryous using		20					
		rnal indexing, Information about various mathematical and							
4	statistical societies, Informa	tion about seminars, conferences and workshops. How to		15					
	read research article (a case study), Methods and processes for solving the problem.								
		List of Textbooks / Reference books							
1		actical Research Methods, New Delhi, UBS Publishers' Distr							
2	Kothari, C.R.,1985, Resear Limited.	ch Methodology-Methods and Techniques, New Delhi,	Wiley	Eas	tern				
3	Kumar, Ranjit, 2005, Resear	ch Methodology-A Step-by-Step Guide for Beginners, (2nd.	ed), S	ingap	ore,				
	<u>-</u>	<u></u>							

Course Code: HUT2102F

	Pearson Education.						
4	Shrivastava, Shenoy& Sharma, Quantitative Techniques for Managerial Decisions, Wiley						
5	Goode W J & Hatt P K, Methods in social research, McGraw Hill						
6	Basic Computer Science and Communication Engineering – R. Rajaram (SCITECH)						
7	Krantz, S. G. A Primer of Mathematical Writing: Second Edition. American Mathematical Society.						
8	Higham, N. J. Handbook of Writing for the Mathematical Sciences. Society for Industrial and Applied						
8	Mathematics.						
9	Christian Heumann, Michael Schomaker, Shalabh, Introduction to Statistics and Data Analysis with						
7	Exercises, Solutions and Applications in R						
10	Brian R. Hunt, Ronald L. Lipsman, Jonathan M. Rosenberg, 2006. A Guide to MATLAB: For Beginners						
10	and Experienced Users						
11	Steven I. Gordon, Brian Guilfoos. 2017. Introduction to Modeling & Simulation with MATLAB® and						
11	Python						
12	Mathematical Computation with Sage by Paul Zimmermann (online book)						
	Course Outcomes (students will be able to)						
CO1	Understand the basics of research methodology						
CO2	Understand the importance and usage of mathematical software in research						
CO3	Understand the basic statistical distribution and basics of testing of hypothesis						
CO4	Get good understanding on various mathematical and statistical journals and indexing						
CO5	Identify directions of research and able to decide on important research questions						

		Mapp	ing of C	ourse O	utcomes	s (COs)	with Pro	gramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	1	1	3	3	2	2	3	2	3	1	1	3
CO2	1	1	3	3	1	1	3	2	3	1	0	3
CO3	1	1	3	3	1	1	3	1	3	1	0	3
CO4	1	1	3	3	1	2	3	3	3	0	0	3
CO5	2	1	3	3	2	3	3	3	3	3	1	3

·	Manning of Course Out Aver (COs) with Programme Specific Outcomes (PSOs)											
Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)  PSO1 PSO2 PSO3 PSO4 PSO5 PSO6												
CO1	0	0	1	3	0	1						
CO2	0	0	1	3	0	1						
CO3	0	0	1	3	0	1						
CO4	0	0	1	3	0	1						
CO5	100	0	1	3	0	1						

	Course Code: MAP 2523	Course Title: Computational Mathematics Lab – I	Credits =								
>	Course Code: WIAF 2525	Course Title: Computational Mathematics Lab – 1	L	T	P						
	Semester: I	Total contact hours: 60	0	0	4						
List of Prerequisite Courses											
	List of C	Courses where this course will be prerequisite									
It is a	foundation course which w	vill be prerequisite for all the courses related to statistic	es ai	nd a	pplied						
mathen	natics.										
	Description of relevance	of this course in the M.Sc. Engineering Mathematics Prog	ram								
This co	ourse will introduce basics of	Python Programming and various numerical methods which	h ar	e use	ful in						

solving differential equations, solving system of linear equations, understanding of machine learning algorithms etc.

	Course Contents (Topics and subtopics)	Hours							
	Module -I (Python Programming)								
1.	Introduction to Python Programming, Python as an advanced scientific calculator, use of math and cmath modules	2							
2	Strings, List, tuples and dictionary data structures in Python, If and else controls and its applications	2							
3	Loops in Python, Creating user defined functions and python modules	4							
4	Vectors and matrix computations in Python using Numpy module	2							
5	Use of SciPy and Sympy Module to solve problems in numerical methods	2							
6	2d and 3d Plotting using Matplotlib	2							
7	Classes in Python with applications	2							
8	Exploring data in Python using Pandas	2							
9	Development of Python Programs for problems in numerical methos of module-II along with exploring error analysis.	15							
	Module -II (Basis of Numerical Methods)								
10	Error Analysis and difference table	2							
	Solution of Algebraic and transcendental equation: Bisection method, Secant method,								
11	Regula-False method, Newton-Raphson method, and convergence criteria for these methods.	4							
	Numerical solution of linear equations: Gauss-Jacobi, Gauss-Seidel iteration,								
12	Successive over relaxation (SOR) and under relaxation method and convergence criteria for these methods.								
13	<b>Interpolations:</b> Lagrange Interpolation, Divided difference, Newton's backward and forward interpolation, Central difference interpolation (Hermite), Cubic Spline.								
14	Numerical differentiation, and integration (Trapezoidal rule, Simpsons 1/3 ,3/8 rules). Gauss quadrature formula								
	Numerical solution of initial value problems (first and higher order ODE): Euler								
	meths, Taylor series method, Runge-Kutta explicit methods (second and forth order),								
15	Predictor—Corrector methods (Adam-Basforth, Adam-Moulton method). Stiff differential	6							
	equations and its solutions with implicit methods, Numerical Stability, Convergence, and								
	truncation Errors for the different methods.								
16	Numerical Solution of boundary value problems using initial value method and Shooting techniques.	3							
	List of Textbooks/ Reference Books								
1.	Dimitrios Mitsotakis, Computational Mathematics: An Introduction to Numerical Scientific Computing with Python, CRC Press, First Ed.	Analysis and							
2	David Beazley, Python Cookbook: Recipes for Mastering Python 3								
	M. K. Jain, S. R. K. Iyengar and R. K. Jain: Numerical methods for scientific an	d engineering							
3	computation, Wiley Eastern Ltd. Third Edition.								
4	Jaan Kiusalaas, Numerical Methods in Engineering with Python, Cambridge University Pro	ess							
5	D.V. Griffiths and I.M. Smith, Numerical Methods for Engineers, Blackwell Scientific Pub								
6	S.D. Conte and C. deBoor, Elementary Numerical Analysis-An Algorithmic Approach, Mc	Graw Hill.							
7	S.C. Chapra, and P.C. Raymond, Numerical Methods for Engineers, Tata Mc Graw Hill.								
8	M.K. Jain: Numerical solution of differential equations, Wiley Eastern, 2nd Ed								
9	Rajesh Kumar Gupta, Numerical Methods Fundamentals and Applications, Cambridge U	Jniv. Press, 1 <sup>s</sup>							
10	Hans Petter Langtangen (auth.)-A Primer on Scientific Programming with Python, Springe	r.							
	Course Outcomes (students will be able to)								
CO1	understand basic of python programming.								

CO2	develop python programmes for problems arising in science and engineering.	
CO3	perform computations with vectors and matrices in Python	
CO4	find numerical solutions of linear and nonlinear equations.	
CO5	solve problems in involving interpolation and its applications	
CO6	model and solve real life problems using ordinary differential equations.	

		Mapp	ing of C	ourse O	utcome	s (COs)	with Pro	ogramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	5	1	5	3	2	1	3	5	1	1	5
CO2	0	5	1	2	1	1	1	1	5	1	1	5
CO3	0	5	1	2	1	1	1	2	5	1	1	5
CO4	0	5	3	2	3	1	2	1	5	1	1	5
CO5	0	5	4	4	4	2	4	1	5	3	1	5
CO6	0	5	3	2	3	1	4	1	5	1	4	5

N	Iapping of Cou	rse Outcomes (	COs) with Prog	ramme Specific	Outcomes (PSC	S)
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6
CO1	0	3	1	3	0	1
CO2	0	3	1	3	0	1
CO3	0	3	1	3	0	1
CO4	0	3	1	3	0	1
CO5	0	3	1	3	0	1
CO6	0	3	1	3	0	1

	Course Code:	Course Title: Elective – I	C	redi	ts = 4
	Course Code:	Course Title: Elective – I	L	T	P
	Semester: I	Total contact hours: 60	4	0	0
Depart	ment will offer elective cours	es. A consolidated list of all the elective subjects is given at the	eno	1.	

SEMESTERII Approve by Academie

Course Code: MAT 2225	Course Titles Differential Equations	Credits = 4			
Course Code: MAT 2235 Semester: II	Course Title: Differential Equations	L	T	P	
Semester: II	Total contact hours: 60	4	0	0	
	List of Prerequisite Courses				

Basic course on Calculus and ordinary differential equations.

# List of Courses where this course will be prerequisite

Advanced Differential Equations (MAT 2233), Mathematical Modelling (MAT 2234), Computational fluid dynamics (MAT 2611)

# Description of relevance of this course in the M.Sc. Engineering Mathematics Program

Ordinary differential equations are in the core of Applied Mathematics and this program emphasize on the applications of mathematics in different branches of science and engineering including industry.

	Course Contents (Topics and subtopics)	Hours						
1	Review of first and second order ODE s Modelling differential equations.	4						
2	Existence and Uniqueness theorems for first order ODEs.	4						
3	Higher Order Linear Equations and linear Systems: fundamental solutions, Wronskian, variation of constants, matrix exponential solution, behaviour of solutions.	16						
4	Boundary Value Problems for Second Order Equations: Green's function, Sturm comparison theorems and oscillations, eigenvalue problems.	10						
5	First order PDEs: Linear, quasi-linear equations-Method of characteristics, Lagrange Methods.	10						
6	Solution of parabolic, elliptic, and hyperbolic equations using variable separable methods.	8						
7	Laplace Transform and Fourier Transform and its application to solve initial value problems and PDEs.	8						
	List of Textbooks/ Reference Books							
1	1 William E. Boyce, Richard C. DiPrima, Elementary Differential Equation, Wiley							
2	E. A. Coddington, An Introduction to Ordinary Differential Equations, PHI							
3	G. F. Simons, S. G. Krantz, Differential Equation, Theory Techniques and Practice Tata M	cGraw-Hill						
4	Zill, Dennis G, A First Course in Differential Equations, Cengage Learning							
5	L.Perko, Differential Equations and Dynamical Systems, 2 <sup>nd</sup> Ed., Springer Verlag.							
6	I. N. Sneddon, Elements of partial differential equations, McGraw-Hill.							
7	W. A Strauss Partial, differential equations, An Introduction, Wiley, John & Sons.							
8	Renardy and Rogers, An introduction to PDE's, Springer-Verlag.							
	Course Outcomes (students will be able to)							
CO1	model real world problems using ordinary and partial differential equation models							
CO2	solve higher order ordinary differential equations using various techniques.							
CO3	investigate the qualitative nature of solutions of ordinary differential equations.							
CO4	solve first order PDEs using various techniques							
CO5	apply various techniques to obtain solutions of heat, wave, and Laplace equations.							

<i>&gt;</i>	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)												
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	
CO1	0	3	0	1	3	3	1	3	3	2	1	3	
CO2	0	3	0	0	1	1	3	0	3	1	0	1	
CO3	2	3	1	1	1	1	3	1	3	0	0	2	
CO4	3	3	1	1	1	1	3	0	3	1	0	2	
CO5	0	3	0	0	1	1	3	1	3	1	1	2	

N	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6						
CO1	3	3	0	0	0	0						
CO2	3	2	0	0	0	0						
CO3	3	2	0	0	0	0						
CO4	3	2	0	0	0	0						
CO5	3	2	0	0	0	0						

	Course Code: MAT 2231	Course Title: Modern Algebra	Credits = 4 L T P					
	Semester: II	Total contact hours: 60	4 0 0					
		List of Prerequisite Courses	9					
NIL			,					
	List of (	Courses where this course will be prerequisite						
Advan	ced Modern Algebra (MAT 26							
	<b>Description of relevance</b>	of this course in the M.Sc. Engineering Mathematics Pro	ogram					
It is a f	*	hematics having various applications in all branches of mat	nematics.					
	Course C	Contents (Topics and subtopics)	Hours					
	Groups, subgroups, cosets,	Lagrange Theorem, Normal subgroups, quotient groups	3.					
1	Focus on symmetric and alternating groups, Symmetry groups Dihedral groups as group							
	of symmetries of a regular n-							
2	-	Direct product of groups, Fundamental theorem for finit	e   8					
	abelian groups (without proo							
3	-	bilizers, applications to the structure of groups, application	s   10					
	to combinatorics.							
4		Integral domains and division rings. Focus on finite fields	5, 10					
		rings, roots and their multiplicities, matrix rings.						
5		Chinese remainder theorem, Euclidean domains, principa	10					
		etorization domains, irreducibility of polynomials.						
_	Extension fields, algebraic extensions, construction of finite fields, roots of polynomials							
6	and splitting fields, constructions with ruler and compass. Polynomial rings and matrix							
	rings over finite fields.	71 ( 0T) (1 1 /D 0						
1	I A C III C	List of Textbooks/ Reference Books						
1	// \ =	Abstract Algebra, 4th Edition, Narosa.						
2		in Abstract Algebra", 7th Ed. Pearson Education.						
3		ote, Abstract Algebra, 2nd Edition, John Wiley.						
5	M. Artin, Algebra, Prentice I							
	G. Santhanam, Algebra, Nar							
6	-	Group Theory: An Expedition with SageMath, Narosa						
CO1	understand basic concepts in	rse Outcomes (students will be able to)						
CO2	investigate basic notions by							
CO2		<u> </u>						
CO3	categorize groups of finite or							
CO4	examine fundamental results in groups, rings and fields investigate properties of rings over finite fields.							
COS	investigate properties of ring	s over mine neigs.						

		Mapp	ing of C	ourse O	utcomes	s (COs)	with Pro	gramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12

CO1	1	1	1	0	2	1	3	0	3	0	0	3
CO2	3	1	1	0	2	2	1	0	3	0	0	3
CO3	3	1	2	0	2	1	2	0	3	0	0	2
CO4	3	1	1	0	2	2	1	0	3	0	1	3
CO5	3	1	1	0	2	1	2	0	3	2	1	1

CO1 CO2	PSO1	PSO2	PSO3	PSO4	PSO5	PSO
CO2	3	0	0	0	0	0
	3	0	0	0	0	0
CO3	3	0	0	0	0	0
CO4	3	0	0	0	0	0
CO5	3	0	0	0	0	0
			COUR	cilon		

	Comme Code MAT 2227	Common Trial of Marsh 1 and 1 and 1	Cred	its = 4			
	Course Code: MAT 2327	Course Title: Machine Learning	LT	P			
	Semester: II	Total contact hours: 60	4	0			
		L'at (P)					
Annlie	ed Linear Algebra (MAT 2201)	List of Prerequisite Courses  On Statistical Computing (MAT 2326), Computational Mathe	matice 1	ah _			
	2523)	, Statistical Computing (WAT 2320), Computational Wattie	manes	Lao –			
(1711)							
	List of C	ourses where this course will be prerequisite					
Deep 1	learning and Artificial intelligen	nce (MAT 2328)		0			
	Description of relevance of	of this course in the M.Sc. Engineering Mathematics Progr	ram 🦳				
		te core of modern computational techniques. This course help					
		atistical concepts behind the machine learning algorithms. St	udents a	ılso ge			
exposi	ure to various challenges in solv	- · · · · · · · · · · · · · · · · · · ·					
		ontents (Topics and subtopics)	Но	urs			
		arning, Distinction between supervised and unsupervised					
		a accuracy, Training Error, Test Error, Bias-variance trade-					
	off, Measuring the quality of	erstanding the concept of model flexibility and prediction					
1		of Training and Test MSE. Case study of linear regression					
•		ression. (Emphasize on understanding the universal patterns	1	2			
	using simulated realizations)						
	_	ining and test error rates, Logistic regression, Linear and					
	quadratic discriminant analys	is					
	Model Selection and Regu	ularization: Multiple Linear Regression, Validation set					
2	approach, Leave-One-Out-Cross-Validation, K-fold cross validation, best subset						
-	selection, Forward Selection, Backward selection, Hybrid selection, shrinkage methods:						
		ampling methods and its application in real data analysis.					
3		Boosting, Random Forests, Gradient Boosting, Adaboost	]	0			
4		Fitting Neural Networks, Selection of number of hidden		8			
	layers, Computational considerational Courses Discriminant Analysis	rsis, Naive Bayes, Support Vector Machines: support vector					
5	classifier, SVM and for regre		1	0			
		cipal Component Analysis, Factor Analysis, Principal					
6		eans clustering, Hierarchical Clustering, Multi-dimensional		_			
	scaling	C,		12			
7	Software Component: R/Pyth	non (Its Implementation will be covered in Computational					
/	Mathematics – II)						
		List of Textbooks/ Reference Books					
1		h Guido, Introduction to Machine Learning with Python: D	avid Ba	rber 1			
	Guide for Data Scientists, O'I						
2	·	with R by Bradley Boehmke and Brandon Greenwell, CRC P					
3		earning with Application in R by James, G., Witten, D.,	Hastie,	T. an			
1	Tibshirani, R.	www. on Statistical Informachy Lower Wassamman					
4		ourse on Statistical Inference by Larry Wasserman.	Trouce	Uesti-			
5	Springer.	Learning by Jerome H. Friedman, Robert Tibshirani, and	rrevor	⊓asti€			
6		n to Machine Learning, The MIT Press, Cambridge.					
U	= -	, Mark A. Hall, Data Mining: Practical Machine Learni	ng Too	ıle an			
7	Techniques by Elsevier	, mark A. Han, Data Minnig. Hactical Machine Lealin	.ng 100	no all			
8		ilistic Perspective (Adaptive Computation and Machine Lear	ning sei	ies) h			

	Kevin P. Murphy.						
	Course Outcomes (students will be able to)						
CO1	understand advantages of machine learning algorithms.						
CO2	apply machine learning techniques to solve regression problems involving real data.						
CO3	apply machine learning techniques to solve classification problems involving real data.						
CO4	apply dimension reduction methods to solve problems involving real data.						
CO5	use software to build machine learning models and interpret the results.						

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	1	3	3	2	1	1	0	3	0	0	3
CO2	0	1	3	3	3	3	3	3	3	2	0	3
CO3	0	1	3	3	3	3	3	3	3	2	0	3
CO4	0	1	3	3	2	3	3	3	3	2	2	3
CO5	0	1	3	3	2	3	3	3	3	2	2	3

N	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)								
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6			
CO1	0	1	3	1	3	0			
CO2	0	1	3	1)	3	0			
CO3	0	1	3	1	3	0			
CO4	0	1	3	(C) 1	3	0			
CO5	0	1	3	0	3	3			

Course Code: MAD 2524	Course Titles Computational Mathematics Lab. II	C	ts = 2					
Course Code: MAP 2524 Course Title: Computational Mathematics Lab – II								
Semester: II	Total contact hours: 60	0	0	4				
AO <sup>y</sup>								
	List of Prerequisite Courses							
C								
List of Courses where this course will be prerequisite								
ced Statistical Computing (MA	AT 2329)							
	List of C	Semester: II Total contact hours: 60  List of Prerequisite Courses	Course Code: MAP 2524   Course Title: Computational Mathematics Lab – II   L  Semester: II   Total contact hours: 60   0    List of Prerequisite Courses  List of Courses where this course will be prerequisite	Semester: II Total contact hours: 60 0 0  List of Prerequisite Courses  List of Courses where this course will be prerequisite				

# Description of relevance of this course in the M.Sc. Engineering Mathematics Program

This M.Sc. program gives special emphasis on the implementation and application of large-scale computational techniques from applied mathematics and statistics. Hence, a good efficiency in mathematical programming is required in the upcoming semesters. Programming lab will give the students exposure to computational mathematics using latest software.

Course Contents (Topics and subtopics)								
Module – I (Basic theory of statistical simulation)								
<i>Y</i>	Simulating Random numbers: Probability Integral transform, Approximating							
	probabilities by means of simulation, Demonstration of Convergence in Probability							
	Using Simulation, Introduction to Monte Carlo Simulation, Demonstration of Weak Law							
1	of Large Numbers, Demonstration of Central Limit Theorem (concepts covered in							
	Statistical Computing), Computing Risk function and comparing risk functions by							
	simulation under different loss functions, Power curves, and comparing testing							
	procedures using power curves.							
2	Statistical analysis of Multiple Linear Regression problem (proofs of unbiasedness,	6						
2	consistency of the estimator), statistical analysis of nonlinear regression models.							
3	Theory of Generalized linear models, estimation, and inference: Poisson regression,	6						

	Logistic regression, Generalized additive models							
4	Multivariate normal distribution and related testing of hypothesis problems	2						
	Module – II (Machine Learning using R/Python)	2						
5	A refresher on R/Python programming	4						
	Building classification models in R/Python using logistic regression, linear discriminant	+ -						
	analysis, quadratic discriminant analysis, checking accuracy using Confusion matrix,							
6	AUC and ROC curves, building classifiers using Naïve Bayes and K-nearest neighbour	6						
	methods, Support vector machines.  Regression problem using R/python: handling problems with qualitative predictors in							
	regression, Interaction between features, understanding the output and interpretation,							
7	regression diagnostics, case studies using real data sets, comparison with k-nearest	6						
	neighbour regression.							
	Model regularization in R/Python: Feature Engineering, Ridge, Lasso, Elastic net, best	10'						
8	subset selection, case studies	4						
	Multivariate methods in R/Python: Principal Component Analysis, Multidimensional	<del>)</del>						
9	scaling, Principal component regression, case studies using real data sets, Clustering	6						
9	methods, matrix completion	0						
	Nonlinear models in R/Python: Nonlinear regression, Regression splines, local							
10	regression, generalized additive models and their applications in solving real life	4						
10	problems.							
11	Building Neural Network models in R/Python and its application to real data analysis	6						
- 11	Data analysis using Tree based methods: Classification trees, regression trees, Bagging,	0						
12	Random Forest and boosting, case studies using real data sets.	4						
	Several case studies from various domains like banking, finance, social sciences,							
13	marketing, biology etc will be covered. Students will do group projects followed by							
13	presentation.							
		<u> </u>						
	List of Textbooks/ Reference Books							
1	Hans Petter Langtangen (auth.)-A Primer on Scientific Programming with Python-S	pringer Berlin						
	Heidelberg.							
2	Reema Thareja, Python Programming: Using Problem Solving Approach.							
3	David Beazley, Python Cookbook: Recipes for Mastering Python 3.							
4	Victor A. Bloomfield, Using R for Numerical Analysis in Science and Engineering, CRC							
5	James, G. Witten, D., Hastie, T. and Tibshirani, R. Introduction to Statistical Learning with	th Applications						
	in R, Springer.							
6	Brian Dennis, The R Student Companion, CRC Press, Taylor and Francis Group.							
7	Garrett Grolemund, Hands-On Programming with R: Write Your Own Functions an	d Simulations,						
	Shroff/O'Reilly.							
8	Laura Chihara and Tim Hesterberg, Mathematical Statistics and Resampling and R.John V							
9	Christian P. Robert and George Casella, Introducing Monte Carlo Methods with R, Spring							
10	Gareth James, Daniela Witten, Trevor Hastie, Robert Tibshirani, Introduction to Statis	stical Learning						
	with Applications in R, Second edition, Springer, 2021							
11	Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, The Elements of Statistical I	Learning, 2003,						
	Springer Publications							
	Course Outcomes (students will be able to)							
CO1	Simulate random numbers from a given probability distribution.							
CO2	Solve the testing problems related to means and variances of the multivariate normal							
	distribution							
CO3	Build classifier to perform prediction and inference tasks using real data sets							
	involving classification problems using software packages							
CO4	Build predictive models using real data sets involving regression problems and							

	perform feature engineering	
CO5	Apply tree-based methods to solve regression and classification problems using real	
CO3	data sets using software packages	
CO6	Train neural network for regression and classification tasks for data analytics	
C06	problems and perform model tuning.	

		Mapp	ing of C	course O	utcome	s (COs)	with Pro	ogramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	1	3	3	1	3	1	3	3	1	0	3
CO2	0	1	3	3	1	3	1	3	3	1	1	73
CO3	0	3	3	3	3	3	3	3	3	1	0	3
CO4	0	0	3	3	4	3	3	3	3	2	1	3
CO5	0	3	3	3	3	3	3	3	3	3	2	3
CO6	0	3	3	3	3	3	3	3	3	3	0	3

N	<b>Iapping of Cou</b>	rse Outcomes (C	COs) with Prog	ramme Specific	Outcomes (PSO	os)
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6
CO1	0	0	3	2	3	0
CO2	0	0	3	2	3	0
CO3	0	0	3	107	3	2
CO4	0	0	3	1	3	2
CO5	0	0	3	O 1	3	1
CO6	0	0	3	1	3	1

	Course Code:	Course Title: Fleetive H	C	ts = 4	
	MATXXXX	Course Title: Elective – II		T	P
	Semester: II	Total contact hours: 60		0	0
Depart	ment will offer electives. A co	nsolidated list of all the elective subjects is given at the end.			

Course Code: MAT 2811	Course Title: On Job Training (OJT)	C	redi	ts = 4
Course Coue: WIAT 2011	Course Title: On Job Training (OJT)	L	T	P
Semester: II	Total contact hours: 60	0	0	6

Candidate will have to undergo a compulsory training with industry as assigned by the department during the semester break. The evaluation will be out of 100 marks. The guidelines, adopted by the Institute has been provided at the end of the document.

Approve by Academic Council on Aug. of 202?

Approve by Academic

# Approve by Academic Council on Aug. of Approve by Academic

	Course Code: MAT 2220	Course Title: Measure, Integration and Functional	C	redit	s = 4
	Course Code: MAT 2229	Analysis	L	T	P
Ī	Semester: III	Total contact hours: 60	4	0	0
	l				-
		List of Prerequisite Courses			
Real ar	nd Complex Analysis (MAT 22	230), Applied Linear Algebra (MAT 2201)			
		Courses where this course will be prerequisite			
Operate	or Theory (MAT 2647)				
		of this course in the M.Sc. Engineering Mathematics Prog			h
		ed and Pure Mathematics. A lot of techniques from Function		b 1.4	7
useful	in differential equations and n	numerical methods. This course strengthens mathematical for	ında	tion	of the
student	ts.				
	•	Contents (Topics and subtopics)		Hou	ırs
1		measure. Lebesgue Measure and its properties. Non-		15	<b>`</b>
		functions and their properties.		- 1.	
2		I convergence theorem, Monotone Convergence theorem,		15	5
	Fatou's Lemma, Dominated				
3	_	nded linear operators and functionals on normed spaces,		12	)
	Banach spaces				· 
		neorem. Zabreiko's lemma for subadditive functionals,			
4		ciple, Closed Graph Theorem, Open Mapping Theorem,		18	<b>,</b>
	Bounded Inverse Theorem as	s consequences of Zabreiko's Lemma.			
		List of Textbooks/ Reference Books			
1		functional Analysis with Applications, John Wiley & Sons, No.	ew Y	ork.	
2	-	alysis, 2ndEdition, New Age International, New Delhi.			
3		nal Analysis for Scientists and Engineers, Springer- Singapor			
4		ar, Functional Analysis—A First Course, Narosa Publishing F	ious	e.	
5		First Course in Functional Analysis, Prentice Hall.			
6 7		l Analysis, Hindustan Book Agency.			
8	H. L. Royden, Real Analysis	leasures and Integration, AMS			
	•	ry and Integration, New Age Publishers, Second Edition			
9	G. De. Darra, Measure Theor	y and integration, new Age rubilshers, second Edition			
	Com	rse Outcomes (students will be able to)			
CO1		of measure as generalization of notion of length.			
CO2		rable functions, and construct non-measurable set			
CO3	•	otone, dominated convergence theorems			
		operators on normed spaces and give an example of			
CO4	noncontinuous operator on in				
		Lemma and apply it to prove the major theorems of			
CO5	functional analysis.	Zemme and appropries to prove the imager theorems of			
CO6	compute Hahn Banach exten	sions of linear operators.			
	1	- T			

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)												
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	
CO1	3	1	0	0	2	0	0	0	3	0	0	3	
CO2	3	1	0	0	3	0	1	0	3	1	0	3	
CO3	3	3	1	1	3	0	0	0	3	0	0	3	

CO4	3	1	0	0	2	1	1	0	3	1	0	3
CO5	3	1	0	0	2	1	0	0	3	0	0	3

Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6					
CO1	3	0	0	0	0	0					
CO2	3	0	0	0	0	0					
CO3	3	0	0	0	0	0					
CO4	3	0	0	0	0	0					
CO5	3	0	0	0	0	0					

	Course Code: MAT 2232	Course Titles Ontimization Techniques	C	redit	ts = 4				
	Course Code: MA1 2232	Course Title: Optimization Techniques	L	T	P				
	Semester: III	Semester: III Total contact hours: 60							
		List of Prerequisite Courses							
Applie	ed Linear algebra (MAT 2201)	7 0							
	List of Cou	rses where this course will be prerequisite							
	Description of relevance of t	his course in the M.Sc. Engineering Mathematics Pr	ogram						

# This M.Sc. program gives special emphasis on the implementation and application of large-scale computational techniques from applied mathematics and statistics. Optimization problems are abundant almost in all real-life

problems related to industrial applications.

	Course Contents (Topics and subtopics)	Hours
1	Introduction to Optimization problems and formulations	4
2	<b>One dimensional Optimization:</b> Golden Section method, Fibonacci search Method, Polynomial interpolation method, Iterative methods	8
3	Classical optimization Techniques: Unconstrained optimization, Constrained Optimizations: Penalty methods, Method of Lagrange multiplier, Kuhn-Tucker method	8
4	<b>Linear Programming:</b> Simplex Method, Revised Simplex Method and other advanced Methods, Duality, Dual Simplex Method, Integer Programming Problems	12
5	<b>Unconstrained Optimization Techniques:</b> Direct search methods such as Powel's method, Simplex method, etc	4
6	<b>Gradient Search Methods:</b> Steepest descent method, Conjugate gradient method, Newton's method, Quasi-Newton's method, DFP, BFGS method etc	12
7	Dynamic Programming Problems	4
8	Genetic Algorithms, Simulated Annealing, Ant Colony Optimization	8
	List of Textbooks/ Reference Books	
1	Edvin K. P. Chong & Stanislab H. Zak, An Introduction to Optimization, John Wiley.	
2	Leunberger, Linear and Nonlinear Programming, Springer	
3	Jorge Nocedal, Stephen J. Wright, Numerical Optimization, Springer	
4	S.S. Rao, Engineering Optimization: theory and practices, New Age International Pvt. Ltd,	
5	K. Deb, Optimization for Engineering Design, Prentice Hall, India	
6	L. Davis, Handbook of genetic Algorithm, New York Van Nostrand Reinhold	
7	Z. Michaleuwicz, Genetic Algorithm+Data Structure=Evolution Programme, Springer-Ver	lag

8	R. K. Belew and M. D. Foundations of Genetic Algorithms, Vose, San Francisco, C	CA: Morgan											
8	Kaufmann.												
	Course Outcomes (students will be able to)												
CO1	formulate optimization problems.												
CO2	understand the standard methods to solve unconstrained and constrained optimization												
CO2	problems.												
CO3	understand linear programming problems.												
CO4	solve optimization problems using various algorithms.												
CO5	apply various algorithms in optimization techniques to solve real life problems.												

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	3	3	3	3	0	1	0	3	0	0	3
CO2	0	3	3	3	2	0	2	0	3	0	0	3
CO3	0	2	3	3	2	0	2	0	3	0	0	3
CO4	0	2	3	3	3	1	2	0	3	0	0	3
CO5	0	3	3	3	3	3	3	3	3	3	0	3

	PSO1	PSO2	PSO3	PSO4	PSO5	PSO
CO1	2	3	1	0	0	0
CO2	2	3	1	2	0	0
CO3	2	3	1	2	0	0
CO4	2	3	1	2	0	0
CO5	2	3	1	0	2	2

Course Code: MAT 2328   Semester: III   Total contact hours: 60   4   0   0				C	redit	ts = 4
List of Prerequisite Courses  Statistical Computing (MAT2326), Machine Learning (MAT 2327)  List of Courses where this course will be prerequisite  Description of relevance of this course in the M.Sc. Engineering Mathematics Program  This course gives the students exposure to large scale mathematical computations in solving real life problems.  Course Contents (Topics and subtopics)  Hours  Machine learning basics and introduction to deep learning  Deep Neural networks, Architecture design, backpropagation, and other differentiation algorithms  Regularization for deep learning, Tree based methods and other ensemble models  Course Course for training deep learning models, Approximate second-order methods, algorithm for adaptive learning rates  Convolutional Networks  Recurrent Networks, long short-term memory, optimization for long terms dependencies  Applications of Deep Learning: Computer vision, Speech recognition, Natural language processing  Software Implementation: R/Python/MATLAB  List of Textbooks/ Reference Books  I lan Goodfellow and Yoshua Bengio and Aaron Courville, Deep Learning, MIT Press.  List of Textbooks/ Reference Books  I lan Goodfellow and Yoshua Bengio and Aaron Courville, Deep Learning, MIT Press.  The Elements of Statistical Learning by Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, Springer.  Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach.  Vovidiu Calin, Deep Learning Architectures: A Mathematical Approach.  Kevin P. Murphy, Machine Learning: A Probabilistic Perspective.  John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  Coul understand basic principles of Deep Learning and artificial Intelligence.  understand that mathematical concepts behind deep learning algorithms.  Coulouderstand statistics and optimization principles in deep neural networks.			Course Title: Deep Learning and Artificial Intelligence	L	T	P
List of Courses where this course will be prerequisite   Description of relevance of this course in the M.Sc. Engineering Mathematics Program		Semester: III	Total contact hours: 60	4	0	0
List of Courses where this course will be prerequisite   Description of relevance of this course in the M.Sc. Engineering Mathematics Program			Title CD 114 C			
Description of relevance of this course in the M.Sc. Engineering Mathematics Program   This course gives the students exposure to large scale mathematical computations in solving real life problems.   Course Contents (Topics and subtopics)   Hours	Statisti	ical Computing (MAT2226)		I		
Description of relevance of this course in the M.Sc. Engineering Mathematics Program   This course gives the students exposure to large scale mathematical computations in solving real life problems.   Course Contents (Topics and subtopics)   Hours	Statisti	ical Computing (MA12320),	iviacilille Learning (WAT 2327)			
This course gives the students exposure to large scale mathematical computations in solving real life problems.    Course Contents (Topics and subtopics)   Hours		List of	Courses where this course will be prerequisite			
This course gives the students exposure to large scale mathematical computations in solving real life problems.    Course Contents (Topics and subtopics)   Hours						<b>h</b>
Course Contents (Topics and subtopics)		_				2
1 Machine learning basics and introduction to deep learning 2 Deep Neural networks, Architecture design, backpropagation, and other differentiation algorithms 3 Regularization for deep learning, Tree based methods and other ensemble models 4 Optimization techniques for training deep learning models, Approximate second-order methods, algorithm for adaptive learning rates 5 Convolutional Networks 4 Recurrent Networks, long short-term memory, optimization for long terms dependencies 6 Recurrent Networks, long short-term memory, optimization for long terms dependencies 6 Applications of Deep Learning: Computer vision, Speech recognition, Natural language processing 8 Software Implementation: R/Python/MATLAB 1 Ian Goodfellow and Yoshua Bengio and Aaron Courville, Deep Learning, MIT Press. 2 The Elements of Statistical Learning by Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, Springer. 3 Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach. 4 Ovidiu Calin, Deep Learning Architectures: A Mathematical Approach. 5 Kevin P. Murphy, Machine Learning: A Probabilistic Perspective. 6 John Paul Mueller, Luca Massaron, Deep Learning for Dummies. 7 Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to) CO1 understand basic principles of Deep Learning and artificial Intelligence. understand statistics and optimization principles in deep neural networks. CO2 understand statistics and optimization principles in deep neural networks.	This co			fe pı		
Deep Neural networks, Architecture design, backpropagation, and other differentiation algorithms  Regularization for deep learning, Tree based methods and other ensemble models  Optimization techniques for training deep learning models, Approximate second-order methods, algorithm for adaptive learning rates  Convolutional Networks  Recurrent Networks, long short-term memory, optimization for long terms dependencies  Applications of Deep Learning: Computer vision, Speech recognition, Natural language processing  Software Implementation: R/Python/MATLAB  List of Textbooks/ Reference Books  Ian Goodfellow and Yoshua Bengio and Aaron Courville, Deep Learning, MIT Press.  The Elements of Statistical Learning by Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, Springer.  Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach.  Voidiu Calin, Deep Learning Architectures: A Mathematical Approach.  Kevin P. Murphy, Machine Learning: A Probabilistic Perspective.  John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  Course Outcomes (students will be able to)  course outcomes (students will be able to)  courderstand basic principles of Deep Learning and artificial Intelligence.  understand basic principles of Deep Learning and artificial Intelligence.  understand the mathematical concepts behind deep learning algorithms.					Hot	ırs
algorithms  Regularization for deep learning, Tree based methods and other ensemble models  Optimization techniques for training deep learning models, Approximate second-order methods, algorithm for adaptive learning rates  Convolutional Networks  Recurrent Networks, long short-term memory, optimization for long terms dependencies  Applications of Deep Learning: Computer vision, Speech recognition, Natural language processing  Software Implementation: R/Python/MATLAB  List of Textbooks/, Reference Books  Ian Goodfellow and Yoshua Bengio and Aaron Courville, Deep Learning, MIT Press.  The Elements of Statistical Learning by Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, Springer.  Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach.  Voidiu Calin, Deep Learning Architectures: A Mathematical Approach.  Kevin P. Murphy, Machine Learning: A Probabilistic Perspective.  John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  Venkata Reddy Konaşani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  Course Outcomes (students will be able to)  course outcomes (students will be able to)  course outcomes (students will be applied to maderated the mathematical concepts behind deep learning algorithms.		_		h	6	
Optimization techniques for training deep learning models, Approximate second-order methods, algorithm for adaptive learning rates  Convolutional Networks  Recurrent Networks, long short-term memory, optimization for long terms dependencies  Applications of Deep Learning: Computer vision, Speech recognition, Natural language processing  Software Implementation: R/Python/MATLAB  List of Textbooks/ Reference Books  I Ian Goodfellow and Yoshua Bengio and Aaron Courville, Deep Learning, MIT Press.  The Elements of Statistical Learning by Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, Springer.  Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach.  Ovidiu Calin, Deep Learning Architectures: A Mathematical Approach.  Kevin P. Murphy, Machine Learning: A Probabilistic Perspective.  John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  Col understand basic principles of Deep Learning and artificial Intelligence.  understand the mathematical concepts behind deep learning algorithms.  Col understand statistics and optimization principles in deep neural networks.	2	•	chitecture design, backpropagation, and other differentiation		10	)
methods, algorithm for adaptive learning rates  Convolutional Networks  Recurrent Networks, long short-term memory, optimization for long terms dependencies  Applications of Deep Learning: Computer vision, Speech recognition, Natural language processing  Software Implementation: R/Python/MATLAB  List of Textbooks/ Reference Books  I lan Goodfellow and Yoshua Bengio and Aaron Courville, Deep Learning, MIT Press.  The Elements of Statistical Learning by Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, Springer.  Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach.  Ovidiu Calin, Deep Learning Architectures: A Mathematical Approach.  Kevin P. Murphy, Machine Learning: A Probabilistic Perspective.  John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  Coll understand basic principles of Deep Learning and artificial Intelligence.  understand the mathematical concepts behind deep learning algorithms.  Coll understand statistics and optimization principles in deep neural networks.  Coll apply deep learning algorithms in solving real life problems.	3	Regularization for deep lea	rning, Tree based methods and other ensemble models		6	
5 Convolutional Networks 6 Recurrent Networks, long short-term memory, optimization for long terms dependencies 6 Applications of Deep Learning: Computer vision, Speech recognition, Natural language processing 7 Processing 8 Software Implementation: R/Python/MATLAB 15  List of Textbooks/ Reference Books 1 Ian Goodfellow and Yoshua Bengio and Aaron Courville, Deep Learning, MIT Press. 2 The Elements of Statistical Learning by Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, Springer. 3 Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach. 4 Ovidiu Calin, Deep Learning Architectures: A Mathematical Approach. 5 Kevin P. Murphy, Machine Learning: A Probabilistic Perspective. 6 John Paul Mueller, Luca Massaron, Deep Learning for Dummies. 7 Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to) CO1 understand basic principles of Deep Learning and artificial Intelligence. CO2 understand the mathematical concepts behind deep learning algorithms. CO3 understand statistics and optimization principles in deep neural networks. CO4 apply deep learning algorithms in solving real life problems.	4	-		6		
Applications of Deep Learning: Computer vision, Speech recognition, Natural language processing  8 Software Implementation: R/Python/MATLAB  List of Textbooks/ Reference Books  1 Ian Goodfellow and Yoshua Bengio and Aaron Courville, Deep Learning, MIT Press.  2 The Elements of Statistical Learning by Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, Springer.  3 Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach.  4 Ovidiu Calin, Deep Learning Architectures: A Mathematical Approach.  5 Kevin P. Murphy, Machine Learning: A Probabilistic Perspective.  6 John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  7 Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  CO1 understand basic principles of Deep Learning and artificial Intelligence.  CO2 understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.	5				4	
Applications of Deep Learning: Computer vision, Speech recognition, Natural language processing  8 Software Implementation: R/Python/MATLAB  List of Textbooks/ Reference Books  1 Ian Goodfellow and Yoshua Bengio and Aaron Courville, Deep Learning, MIT Press.  2 The Elements of Statistical Learning by Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, Springer.  3 Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach.  4 Ovidiu Calin, Deep Learning Architectures: A Mathematical Approach.  5 Kevin P. Murphy, Machine Learning: A Probabilistic Perspective.  6 John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  7 Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  CO1 understand basic principles of Deep Learning and artificial Intelligence.  CO2 understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.	6	Recurrent Networks, long s	short-term memory, optimization for long terms dependencies		6	
processing  8 Software Implementation: R/Python/MATLAB  List of Textbooks/ Reference Books  1 Ian Goodfellow and Yoshua Bengio and Aaron Courville, Deep Learning, MIT Press.  2 The Elements of Statistical Learning by Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, Springer.  3 Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach.  4 Ovidiu Calin, Deep Learning Architectures: A Mathematical Approach.  5 Kevin P. Murphy, Machine Learning: A Probabilistic Perspective.  6 John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  7 Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  CO1 understand basic principles of Deep Learning and artificial Intelligence.  CO2 understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.	7	_				
Software Implementation: R/Python/MATLAB					7	
1 Ian Goodfellow and Yoshua Bengio and Aaron Courville, Deep Learning, MIT Press. 2 The Elements of Statistical Learning by Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, Springer. 3 Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach. 4 Ovidiu Calin, Deep Learning Architectures: A Mathematical Approach. 5 Kevin P. Murphy, Machine Learning: A Probabilistic Perspective. 6 John Paul Mueller, Luca Massaron, Deep Learning for Dummies. 7 Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to) CO1 understand basic principles of Deep Learning and artificial Intelligence. CO2 understand statistics and optimization principles in deep neural networks. CO3 apply deep learning algorithms in solving real life problems.	8		R/Python/MATLAB		15	5
The Elements of Statistical Learning by Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie, Springer.  Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach.  Ovidiu Calin, Deep Learning Architectures: A Mathematical Approach.  Kevin P. Murphy, Machine Learning: A Probabilistic Perspective.  John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  CO1 understand basic principles of Deep Learning and artificial Intelligence.  Understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.		-	List of Textbooks/ Reference Books			
Springer.  Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach.  Ovidiu Calin, Deep Learning Architectures: A Mathematical Approach.  Kevin P. Murphy, Machine Learning: A Probabilistic Perspective.  John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  CO1 understand basic principles of Deep Learning and artificial Intelligence.  CO2 understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.	1	Ian Goodfellow and Yoshu	a Bengio and Aaron Courville, Deep Learning, MIT Press.			
Josh Patterson, Adam Gibson, Deep Learning: A Practitioner's Approach.  Ovidiu Calin, Deep Learning Architectures: A Mathematical Approach.  Kevin P. Murphy, Machine Learning: A Probabilistic Perspective.  John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  CO1 understand basic principles of Deep Learning and artificial Intelligence.  CO2 understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.	2		l Learning by Jerome H. Friedman, Robert Tibshirani, and	Tre	vor I	Hastie,
4 Ovidiu Calin, Deep Learning Architectures: A Mathematical Approach.  5 Kevin P. Murphy, Machine Learning: A Probabilistic Perspective.  6 John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  7 Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  CO1 understand basic principles of Deep Learning and artificial Intelligence.  CO2 understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.	3		on Deen Learning: A Practitioner's Approach			
5 Kevin P. Murphy, Machine Learning: A Probabilistic Perspective. 6 John Paul Mueller, Luca Massaron, Deep Learning for Dummies. 7 Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  CO1 understand basic principles of Deep Learning and artificial Intelligence.  CO2 understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.			7 7 7			
6 John Paul Mueller, Luca Massaron, Deep Learning for Dummies.  7 Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  CO1 understand basic principles of Deep Learning and artificial Intelligence.  CO2 understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.						
Venkata Reddy Konasani, Shailendra Kadre, Machine Learning and Deep Learning Using Python and TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  CO1 understand basic principles of Deep Learning and artificial Intelligence.  CO2 understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.						
TensorFlow, Mc Graw Hill.  Course Outcomes (students will be able to)  CO1 understand basic principles of Deep Learning and artificial Intelligence.  CO2 understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.			<u> </u>	ing I	ytho	n and
CO1 understand basic principles of Deep Learning and artificial Intelligence.  CO2 understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.		Y ' '		0	<i>J</i> · · ·	
CO2 understand the mathematical concepts behind deep learning algorithms.  CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.		Course Out	comes (students will be able to)			
CO3 understand statistics and optimization principles in deep neural networks.  CO4 apply deep learning algorithms in solving real life problems.	CO1	understand basic principles	of Deep Learning and artificial Intelligence.			
CO4 apply deep learning algorithms in solving real life problems.	CO2	understand the mathematic	al concepts behind deep learning algorithms.			
	CO3	understand statistics and op	otimization principles in deep neural networks.			
CO5 apply Deep Learning Algorithms using R or Python.	CO4	apply deep learning algorit	hms in solving real life problems.			
	CO5	apply Deep Learning Algor	rithms using R or Python.			

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)												
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	
CO1	0	1	3	3	0	2	2	3	3	1	0	3	
CO2	0	1	3	3	2	3	2	3	3	2	0	3	
CO3	0	2	3	3	2	3	3	3	3	2	3	3	
CO4	0	2	3	3	2	3	3	3	3	3	3	3	
CO5	0	2	3	3	1	3	3	3	3	3	3	3	

Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6					
CO1	0	1	3	1	3	0					
CO2	0	1	3	1	3	0					
CO3	0	1	3	1	3	0					
CO4	0	1	3	1	3	0					
CO5	0	1	3	0	3	3					

Course Code: MAP 2525	Course Title: Computational Mathematics Lab-III			ts = 2	
	Course Code. WIAI 2323	Course True. Computational Wathematics Lab-111	L	T	P
	Semester: III	Total contact hours: 60	0	0	4

# **List of Prerequisite Courses**

Differential Equations (MAT 2235), Computational Mathematics Lab – I (MAP 2523), Computational Mathematics Lab – II (MAP 2524)

# List of Courses where this course will be prerequisite

It is a foundation course which will be prerequisite for all the courses related to statistics and applied mathematics.

# Description of relevance of this course in the M.Sc. Engineering Mathematics Program

This advanced course covers the MATLAB programming language and its applications to solve scientific and engineering problems as an application to ODE and PDE

	Course Contents (Topics and subtopics)	Hours
	Module -I (MALTAB: As a computational Tool)	
1.	Defining vectors and matrices and matrix computations, Fundamental programming structures (if statements, for, while loops), Creating user defined functions, File processing, Plotting 2d and 3d graphics in various formats.	6
2	Development of MATLAB programmes for problems in Numerical Analysis with error analysis. Examples arising from some engineering application may be introduced.	6
3	Numerical solution of initial and boundary value ODE in MATLAB	4
4	Numerical solution of standard partial differential equation using MATLAB	6
5	Development of MATLAB Programmes to solve problems involving Laplace and Fourier Transforms	6
6	A group projects in a group of 3-4 students may be assigned. Projects may be selected from [Danaila et al.]	12
	Module -II (Numerical Solution of PDE and Integral Transforms)	
7	Numerical Solutions of PDE's: Numerical Solution of partial differential equations (parabolic and hyperbolic) using explicit and implicit finite difference methods, Numerical stability for explicit and implicit method. Solution of elliptic equation using finite difference methods, Collocation and Galerkin methods, Methods of finite residuals, Finite element formulation for the solution of ODE and PDE, Calculation of element matrices, assembly, and solution of linear equations.	12
8	Introduction of standard integral transform and Applications	8

	List of Textbooks/ Reference Books
1.	Dingyü Xue, Yang Quan Chen, Scientific Computing with MATLAB, Second Edition, 2021 by
	Chapman & Hall
2	C. F. Van Loan and KY. D. Fan, Insight, Through Computing: A MATLAB Introduction to
	Computational Science and Engineering, SIAM Publication, 2009
3	Eihab B. M. Bashier, Practical Numerical and Scientific Computing with MATLAB and Python, CRC
3	Press, 2020
4	Ionut Danaila, Pascal Joly, Sidi Mahmoud Kaber and Marie Poste, An Introduction to Scientific
7	Computing: Twelve Computational Projects Solved with MATLAB, Springer 2006.
5	Dingyü Xue, Differential Equation Solutions with MATLAB, De Gruyter, 2020
6	Sudhakar Nair, Advanced Topics in Applied Mathematics for Engg. & Physical Science, 1st edition,
0	Cambridge University Press
7	Larry C. Andrews Bhimsen, K. Shivamogga, Integral Transforms for Engineers, SPIE Optical
,	Engineering Press
	Course Outcomes (students will be able to)
CO1	understand the basics of MATLAB programming.
CO2	develop MATLAB programmes to solve problems arising in science and engineering.
CO3	develop MATLAB Programmes for numerical solutions of ODE and PDE
CO4	Perform convergence analysis of numerical method for of PDE
CO5	Develop understanding of Laplace and Fourier Transforms and their applications.
CO6	model and solve real life problems and solve it using MATLAB.

		Mapp	ing of C	Course O	utcomes	s (COs)	with Pro	ogramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	3	0	3	2	• 2	0	1	3	0	0	3
CO2	0	3	0	1	0	2	0	1	3	0	0	3
CO3	0	3	0	1	0	2	0	1	3	0	0	3
CO4	0	3	1	1	1	2	2	0	3	0	0	3
CO5	0	3	2	2	2	2	2	0	3	1	0	3
CO6	0	3	0	1	0	3	0	2	3	3	3	3

M	<b>Lapping of Cou</b>	rse Outcomes (C	COs) with Prog	ramme Specific	<b>Outcomes (PSC</b>	<b>)</b> s)
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6
CO1	100	1	0	3	0	0
CO2	0	1	0	3	0	0
CO3	0	1	0	3	0	0
CO4	3	1	0	3	0	0
CO5	0	2	0	3	0	0
CO6	0	3	0	3	3	1

	Course Code:	Course Title: Elective – III	C	ts = 4	
	MATXXXX	Course Title: Elective – III		T	P
	Semester: III	Total contact hours: 60	4	0	0
Depart	ment will offer elective cours	es. A consolidated list of all the elective subjects is given at the	e end	1.	

Course Code: MAP 2704		Cr	edit	s = 4
Course Code: MAP 2704	Course Title: Research Project			P
Semester: III	Total contact hours: 60	0	0	8

This would be concerned with the continuation of the research project executed in the third semester and the exact work plan will be decided in consultation with the project guide. A suitable combination of the marks for report and presentation will be considered for the final evaluation as per the Institute evaluation policy.

# **Suggested Marks distribution:**

Approve by Academic Council on Aug.

Approve by Academic Internal Marks (40) + Final Presentation (20) + Report (20) + Overall (20) = Total (100)

SEMESTER IV

	G G 1 371 massa		C	redit	ts = 4
	Course Code: MAT 2233	Course Title: Advanced Differential Equations	L	T	P
	Semester: IV	Total contact hours: 60	4	0	0
D. CC	.: 1 (MATE 2025)	List of Prerequisite Courses	1		
Differ	ential equations (MAT 2235)				
	List of C	Courses where this course will be prerequisite			
	List of C	courses where this course will be prerequisite			
	Description of relevance	of this course in the M.Sc. Engineering Mathematics Prog	ram		1
This s		stand the nature of fluid flows and mathematical modelling of			mas
	er phenomena	č	_(	V	)
	Course C	Contents (Topics and subtopics)		Hot	ırs
	Review of solution methods	s for first order as well as second order equations. Power			
1	Series methods for solution	of second order differential equations. Regular singular		12	)
1	points. Solution of Legendre	e and Bessel's equation with properties of Bessel functions		12	2
	and Legendre polynomials.				
		Order Partial Differential Equations, normal forms and			_
2		Boundary Value Problems: Lagrange-Green's identity and		12	2
		ds. Stability theory, energy conservation and dispersion.			
2		ue property, weak and strong maximum principle, Green's		1.0	,
3	method.	Dirichlet's principle, existence of solution using Perron's		12	2
		problem, fundamental solution, weak and strong maximum			
4	principle and uniqueness resu			12	2
		, D'Alembert's method, method of spherical means and		1.0	
5	Duhamel's principle.			12	2
		List of Textbooks/ Reference Books			
1	Renardy and Rogers, An intr	oduction to PDE's, Springer-Verlag.			
2		ntial equations, An Introduction, Wiley, John & Sons.			
3		dvanced Engineering Mathematics, Jones & Bartlett.			
4	L.C. Evans, Partial differenti	· · · · ·			
5	V / *	artial differential equations, McGraw-Hill.			
6		s, Numerical solution of partial differential equations, Cambridae			Edn.
7		ion of partial differential equations, finite difference methods,	Oxi	ord.	
8	77	to Finite Element Methods, McGraw-Hill.	.4- 1	N.	V = .1
9	NY: Clarendon Press.	tion of partial differential Equations: Finite difference method	us, I	new	I OT
		ions and Dynamical Systems, Texts in Applied Mathematic	rs V	7o1 7	7 2n
10	Edition, Springer Verlag, Ne		, ,	01. /	, 211
11		rential Equations, Birkhauser, 1995.			
12	7 /	Equations, 3rd Edition, Narosa, 1979.			
12		tial Equations of Applied Mathematics, 2nd Edition, John W	iley	and	Sons
13	1989.				
•		rse Outcomes (students will be able to)			
CO1		to solve partial differential equations.			
CO2	find numerical solutions of p	-			
CO3	implement algorithms to solv				
CO4		rical solutions of differential equations.	<u> </u>		
CO5	model and solve real life pro	blems using partial differential equations.			

Mapping of Course Outcomes (COs) with Programme Outcomes (POs)												
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	3	0	0	1	1	1	1	3	0	0	3
CO2	2	3	0	1	0	0	0	0	3	0	0	3
CO3	2	3	0	1	0	0	0	1	3	0	0	3
CO4	3	3	0	0	1	0	2	0	3	0	0	3
CO5	0	3	2	2	2	3	2	0	3	2	1	3

Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)							
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6	
CO1	3	0	0	0	0	0	
CO2	3	0	0	0	0	0	
CO3	3	0	0	0	0	0	
CO4	3	0	0	0	0	0	
CO5	3	3	0	0	0.0	0	

	Course Code: MAT 2329	Course Titles Advanced Statistical Computing	Credits = 4				
	Course Code: WIA 1 2329	Course Title: Advanced Statistical Computing		T	P		
	Semester: IV	Semester: IV Total contact hours: 60					
	List of Prerequisite Courses						
Proba	ability Theory (MAT 2321), Stat	istical Inference (MAT 2322), Programming Lab (MAP 2521	l)				
	List of C	ourses where this course will be prerequisite	•				
		ACY					
	Description of relevance	of this course in the M.Sc. Engineering Mathematics Prog	ram				
With	an enormous increase of the	large-scale computational methods in science and engine	eerin	ıg, a	pplied		

With an enormous increase of the large-scale computational methods in science and engineering, applied mathematicians must get exposure to various statistical methods. This course aims to give the students exposure to computer intensive statistical methods. It also enables students to understand various simulation methods and Monte carlo techniques which are in the core of application of mathematics to solve real life problems.

Course Contents (Topics and subtopics)				
1	Estimation of cumulative distribution function and statistical functionals	6		
2	Approximation of the distribution of nonlinear functions of random variables and functions of random sample: (Central Limit Theorem and First order and second order Delta method, Extension to multivariate delta method)	6		
3	Random variable generation: Simulation of Random numbers following some specific distribution; Probability Integral transform; Accept/Reject algorithm; Metropolis algorithm, Gibbs sampler	8		
4	Monte Carlo Integration, Importance Sampling, Variance reduction, Riemann Approximations, Laplace Approximations, Saddle point approximation, Acceleration using Antithetic variables, control variates and conditional expectations, Statistical simulation using R	10		
5	Bootstrap methods: Bootstrap variance estimation, Bootstrap confidence intervals, Jacknife.	6		
6	Elements of Bayesian inference: Bayesian philosophy, Prior distribution, posterior distribution, computing posterior point estimate, conjugate prior distribution, Jeffrey's	12		

	prior, multi-parameter problems and Bayesian testing, large sample properties of Bayes									
	estimators (emphasis on real data problems and use of packages in R or Python for	•								
	Bayesian inference)									
	Nonparametric curve estimation: Histogram estimator, Kernel density estimation, bias-									
7	variance trade-off, smoothing using orthogonal functions: density estimation and	12								
	regression problems									
	List of Textbooks/ Reference Books									
1	Larry Wasserman, All of Statistics: A concise course in statistical inference.									
2	Daniel Sabanés Bové and Leonhard Held, Applied Statistical Inference: Likelihood and I	Bayes, Springer.								
3	Christian P. Robert George Casella, Monte Carlo Statistical Methods, Springer.	0-								
4	Eric A. Suess, Bruce E. Trumbo, Introduction to Probability Simulation and Gibbs Sa	ampling with R,								
-	Springer.									
5	ames R. Thompson, Simulation A Modeler's Approach, John Wiley & Sons, Inc.									
6	Reuven Y. Rubinstein, Dirk P. Kroese, Simulation and the Monte Carle method, John Wiley & Sons,									
	Inc.									
7	Christian P Robert and George Casella, Introducing Monte Carlo Methods with R, Spring	ger								
8	Larry A. Wasserman, All of Nonparametric Statistics, Springer									
9	R. A. Thisted, Elements of Statistical Computing. Taylor and Francis									
	Course Outcomes (students will be able to)									
CO1	approximate the distribution of nonlinear functions of random variables using large sample theory.									
	simulate random numbers from some statistical distribution using different									
CO2	algorithms.									
CO3	apply Monte Carlo simulation to estimate model parameters and draw inference.									
CO4	understand basic principles of Bayesian statistics and apply them in parameter									
CO4	estimation problems.									
CO5	apply resampling methods to approximate confidence intervals and variance of									
CO3	estimators.									
CO6	apply nonparametric statistical methods to solve real life data analysis problems									

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)												
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	
CO1	0	0	3	3	1	1	1	0	3	0	0	3	
CO2	0	1	3	3	2	1	2	0	3	1	0	3	
CO3	0	1	3	3	1	1	1	0	3	0	0	3	
CO4	0	(0)	3	3	2	1	3	2	3	0	1	3	
CO5	0	0	3	3	1	2	3	2	3	2	1	3	
CO6	0	0	3	3	2	2	3	3	3	3	2	3	

N.	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
<i>\</i>	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6						
CO1	0	0	3	0	0	1						
CO2	0	0	3	0	0	1						
CO3	0	0	3	0	0	1						
CO4	0	0	3	0	0	1						
CO5	0	0	3	0	0	1						
CO6	0	0	3	0	1	1						

	Course Code: MAP 2705	Course Title: Research Project	Credits = 6	İ
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		L	T	P
Semester: IV	Total contact hours: 80	0	0	12

This would be concerned with the continuation of the research project executed in the third semester and the exact work plan will be decided in consultation with the project guide. A suitable combination of the marks for report and presentation will be considered for the final evaluation as per the institute policy.

### **Suggested Marks distribution:**

Internal Marks (40) + Final Presentation (20) + Report (20) + Overall (20) = Total (100)

	Course Code: MAT 2234	Course Title: Mathematical Modelling	Cr	edits T	= 4 P				
	Semester: IV	Total contact hours: 60	4	0	0				
				I					
		List of Prerequisite Courses							
Ordina	ry differential equations (M	IAT 2221), Partial differential equations (MAT 2222),							
Compu	utational Mathematics Lab – I	(MAT 2523)							
	List of C	Courses where this course will be prerequisite							
NIL		, O <sup>y</sup>							
	_	of this course in the M.Sc. Engineering Mathematics Progr							
		oply the theory of ordinary and partial differential equations to	solv	e real	life				
proble	ms arising from engineering, b								
		Contents (Topics and subtopics)	]	Hour	S				
		al modelling using linear and nonlinear discrete dynamical							
1	systems: qualitative analysis of discrete dynamical systems, One dimensional map, two dimensional maps, Lyapunov exponents and chaotic attractor, example from other branches of science.								
	Qualitative analysis of mathematical models governed by differential equations: Planar								
	•	Eigenvectors defining stable and unstable manifolds, Phase							
2		Hartman's theorem, Construction of phase plane diagram,	8						
	Lyapunov functions	traitman's theorem, construction of phase plane diagram,							
	· -	natical models: Equilibrium points and their classifications,							
		stability. Limit cycles: Existence and uniqueness of limit							
3		y of limit cycles, Poincare- Bendixson theorem, worked	8						
	examples from ecology, dise								
4	- 2 \	eory and applications to analyse mathematical models:		10					
4	different types of bifurcation	s and their analysis using computational software tools		10					
5	Applications of Stochastic n	nodels in modelling real life problems: Simulation, analysis		10					
3	and inference from real data.			10					
Κ,	Mathematical Modelling pro	ojects using computational tools like MATLAB/R/Python.							
6	1	ematical models in fisheries management, traffic dynamics,		16					
O	Predator prey systems, age-	structured models in biology, spatial spread of population,		10					
	etc.								
	I a u b	List of Textbooks/ Reference Books							
1		hematical Modelling: Models, Analysis and Applications, So	econd	1 Edi	tion,				
	CRC Press	1.10							
2		nical Systems with Applications using MATLAB. Springer.							
3		ements of Applied Bifurcation Theory, Second Edition, Spring	ger.						
4	L.Perko, Differential Equation	ons and Dynamical Systems, Vol. 7, 2 <sup>nd</sup> Ed., Springer Verlag.							

5	Reinhard Illner, C. Sean Bohun, Samantha McCollum, Thea Van Roode, 2005, Mathemati	ical Modelling:							
3	A Case studies approach, American Mathematical Society.								
6	James T Sandefur, Discrete dynamical systems Theory and applications, Clarendon press.								
7	M W Hirsch and S Smale - Differential Equations, Dynamical Systems, Academic.								
8	R. Clark Robinson. An Introduction to Dynamical Systems Continuous and Discrete, S	Second edition.							
0	American Mathematical Society, Rhode Island.								
9	Rudiger Seydel, Practical Bifurcation and Stability analysis. Springer (3rd Ed).								
10	Alligood, Sauer, and Yorke. Chaos: An Introduction to Dynamical Systems. Springer, Springer-Verlag								
10	New York.								
		0-							
	Course Outcomes (students will be able to)	0.2							
CO1	Construct mathematical models for real life problems	-0							
CO2	Analyse the qualitative features of mathematical models using techniques from								
CO2	dynamical systems								
CO3	Perform local and global bifurcation analysis for nonlinear systems.	7							
CO4	Use symbolic mathematical software to analyse the mathematical models								
CO5	Construct and analyse stochastic models for solving real life problems.								
CO6	Construct and analyse mathematical models using partial differential equations for real								
100	life problems								

		Mapp	ing of C	ourse O	utcomes	s (COs)	with Pro	gramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	3	0	0	2	2	2	1	3	1	1	3
CO2	0	3	0	0	1	1	3	0	3	1	0	3
CO3	0	3	0	1	1	1	3	0	3	1	0	3
CO4	0	3	0	0	1	3	2	0	3	0	1	3
CO5	0	3	0	0	1	<b>3</b>	3	3	3	3	2	3
CO6	0	3	0	0	2	3	1	3	3	2	2	3

N	<b>Sapping of Cour</b>	rse Outcomes (C	COs) with Progr	amme Specific	Outcomes (PSO	(s)
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6
CO1	1	3	0	0	0	0
CO2	1 (	3	0	0	0	0
CO3	<b>1</b>	3	0	0	0	0
CO4	2,1	3	0	0	0	0
CO5	0	3	1	0	0	0
CO6	0	3	0	0	0	1

Approve by Academic Council on Arth **Detailed Syllabus of Electives Courses** 

	Course Code: MAT 2651	Course Titles Cweek Theory	Cr	edits	= 4				
	Course Code: MAT 2651	Course Title: Graph Theory	L	T	P				
	Elective	Total contact hours: 60	4	0	0				
		I '4 (P)							
NIL		List of Prerequisite Courses							
NIL									
	List of C	Courses where this course will be prerequisite							
NIL		* *							
	Description of relevance	of this course in the M.Sc. Engineering Mathematics Progr	am	1					
This is	an elective course to give the	students an exposure of mathematical foundations of graphs	and	netw	orks				
which	have immense applications in s								
		ontents (Topics and subtopics)	]	Hou	`S				
1	_	morphism, sub graphs, matrix representations, degree,		6					
	operations on graphs, degree	sequences. est paths: Walks, trails, paths, connected graphs, distance,							
2	cut vertices, cut-edges, block		8						
3	_	nber of trees, minimum, spanning trees.		6					
4		partite graphs, line graphs, chordal graphs		6					
5	Eulerian graphs: Characterization, Fleury's algorithm, Chinese-postman-problem								
6		Iamilton graphs: Necessary conditions and sufficient conditions							
	Independent sets and cliqu	nes, coverings, matching: Basic equations, matching in			-				
7	bipartite graphs, Halls Theor	em, perfect matching, defect form of Halls Theorem, greedy		10					
	and approximation algorithm								
8	_	atic number and cliques, greedy colouring algorithm,		10					
	colouring of chordal graphs,								
9	graphs, Hamilton directed graphs	e, in-degree, connectivity, orientation, Eulerian directed		6					
	graphs, Hammon directed gr	List of Textbooks/ References							
	Bondy and U.S.R.Murty: Gr	aph Theory and Applications (Freely downloadable from Bo	ndv's	weh	site:				
1	Google-Bondy).	upi 11001) und 12ppirounono (11001) do minouduoto 110111 20	1100	, ,, •	,				
2		raph Theory, Prentice-Hall of India/Pearson.							
3	J.A.Bondy and U.S.R.Murty:	Graph Theory, Springer.							
4	R.Diestel: Graph Theory, Spr								
5		nond Greenlaw, Graph Theory: Modeling, Applications, an	d Al	gorit	hms,				
	Pearson.	4							
6	A 1	than, A textbook of Graph theory. Second edition. Springer.	in a	Com					
7	Limited.	g, Introduction to Graph Theory. Tata McGraw-Hill Publish	iiiig	Com	pany				
		rse Outcomes (students will be able to)							
CO1	describe important classes of	· · · · · · · · · · · · · · · · · · ·							
	explain fundamental theorems on trees matchings connectivity colorings plane and								
CO2	hamiltonian graphs.								
CO3		of trees and illustrate their applications.							
CO4	describe and apply some basis								
CO5	apply graphs as a tool to mod	lel real-life problems.							

# **Mapping of Course Outcomes (COs) with Programme Outcomes (POs)**

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	3	2	3	2	2	2	0	3	0	0	3
CO2	3	3	3	3	0	0	1	0	3	0	0	3
CO3	3	1	2	2	0	0	2	0	3	0	0	3
CO4	1	2	2	3	1	3	2	0	3	1	1	3
CO5	2	2	2	3	1	3	3	0	3	2	1	3

M	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6						
CO1												
CO2												
CO3												
CO4						"						
CO5												

	1			nodii	ts = 4	
	Course Code: MAT 2612	Course Title: Combinatorics		T	P	
	TH. 41	T 11 60	L			
	Elective	Total contact hours: 60	4	0	0	
		List of Prerequisite Courses				
NIL						
	List of C	Courses where this course will be prerequisite				
		of this course in the M.Sc. Engineering Mathematics Progr				
	-	y mathematical foundation and exposure to problems related	to a	pplic	ations	
of disc	rete mathematics in different d					
	VY	ontents (Topics and subtopics)	Hours			
1		efficients, and important identities	4			
2	Recurrences, Fibonacci numl		3			
3		utations, Stirling numbers of both kinds		5		
4	Set Partitions: Exponential matrices	Generating function, Dobinski's formula, orthogonality of		4		
5	Integer Partitions: Euler's pentagonal Number theorem	identity, conjugate partitions, bijective proofs, Euler's		4		
6	Generating functions, ording generating functions, expone	ary and exponential, examples of OGFS, composition of ntial formula for EGFS.		5		
7		, distances, Adjacency matrix of graphs, distance matrix of Theorem, Counting Spanning trees, Matrix Tree theorem,		20	)	
8	Exploration of concepts in co	ombinatorics and graphs theory using Sagemath		15	5	
	- *	List of Textbooks/ Reference Books				
1	Miklos Bona, Introduction to	Enumerative Combinatorics, McGraw-Hill.				
2		Combinatorics, World Scientific.				
3		onal Mathematics with SageMath (free online on sagemath.or	rg).			
4	M. Aigner, A Course in Enui		3/-			
	6,					

5	C. Berge. Principles of Combinatorics. Academic Press.								
6	J. M. Harris, J. L. Hirst, M. J. Mossinghoff, Combinatorics and Graph Theory, Springer.								
7	Istvan Mezo, Combinatorics and number theory of counting sequences, CRC Press.								
	Course Outcomes (students will be able to)								
CO1	understand fundamental mathematical objects such as sets, functions and								
COI	permutations.								
CO2	solve problems involving various counting principles.								
CO3	apply combinatorial ideas to practical problems.								
CO4	understand and use idea of modelling problems using Graph Theory.								
CO5	solve problems in combinatorics and graph theory using SageMath.	C							

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)												
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	
CO1	3	0	1	1	1	1	1	0	3	0	1	3	
CO2	3	1	1	1	0	3	1	1	3	2	0	3	
CO3	0	1	1	1	0	0	3	0	3	0.	0	3	
CO4	0	1	1	1	2	2	1	1	3	79	0	3	
CO5	0	3	3	3	2	2	1	0	3	2	1	3	

N	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6						
CO1				C								
CO2				>								
CO3			. 0									
CO4												
CO5												

	107	<del>_</del>						
	Course Code: MAT 2606	Course Title: Financial Mathematics	C	redit	s = 4			
	Course Coue. WAT 2000	Course Title. Financial Wathematics	L	T	P			
	Elective	Total contact hours: 60	4	0	0			
	1							
	List of	Prerequisite Courses						
	4							
	List of Courses who	ere this course will be prerequisite						
Advan	ced Mathematical Finance (MAT 2610)							
	Description of relevance of this cour	se in the M.Sc. Engineering Mathematics Progra	am					
This c	course will provide a basic introduction	n to financial markets and illustrate application	ns (	of se	veral			
mather	natical models in financial markets.							
	Course Contents (To	opics and subtopics)		Hou	rs			
	The Time Value of Money: Compound	nterest with fractional compounding, NPV, IRR,						
1	and Descartes's Rule of Signs, Annuity	and amortization theory, The Dividend Discount		8				
	Model, Valuation of Stocks, Valuation of	f bonds						
2	Portfolio Theory: Markowitz portfoli	o model, Two-security portfolio, N-security		8				
2	portfolio, Investor utility, Diversification and the uniform Dirichlet distribution							
3	Capital Market Theory and Portfolio Risk Measures: The Capital Market Line, The							
3	CAPM Theorem, The Security Market I	12						
4	Modeling the Future Value of Risky Sec	urities: Binomial trees, Continuous-time limit of		8				

	the CRR tree, Stochastic process: Brownian motion and geometric Brownian motion,							
	Itô's formula.							
	Forwards, Futures, and Options: No arbitrage and the Law of One Price, Forwards,							
5	Futures, Option type, style, and payoff, Put-Call Parity for European options, Put-Call	12						
	Parity bounds for American options							
	The Black-Scholes-Merton Model: Black-Scholes-Merton (BSM) formula, Partial							
6	differential equation approach to the BSM formula: the BSM Partial differential equation	12						
0	Continuous-time, risk-neutral approach to the BSM formula, Binomial-tree approach to	12						
	the BSM formula, Delta hedging, Implied volatility.							
	List of Textbooks/ Reference Books							
1	S.M. Ross, An introduction to Mathematical Finance, Cambridge University Press.	0,3						
2	A. J. Prakash, R. M. Bear, K. Dandapani, G.L. Gahi, T.E. Pactwa and A.M. Parchiga	ari, The return						
2	Generating Models in Global Finance, Pergamon Press.							
3	J. Hull, Options, Futures, and Other Derivatives, Pearson Prentice Hall, Upper Saddle Rive	er.						
4	S. M. Ross, Applied Probability: Models with Optimization Applications, Holdenday.							
5	S. Roman, Introduction to the Mathematics of Finance Springer, New York.							
	Course Outcomes (students will be able to)							
CO1	Understand basic idea of different financial instruments							
CO2	Understand various concepts related to portfolio theory.							
CO3	Model financial instruments using stochastic processes and Ito formula							
CO4	Apply probability concepts for pricing options, future etc.							
CO5	Apply Black-Scholes model for option pricing							

		Mapp	ing of C	ourse O	utcomes	s (COs)	with Pro	ogramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	2	1	1	1	1	1	0	3	0	0	3
CO2	0	2	1	2	0.	0	2	0	3	0	0	3
CO3	1	3	2	1	1	0	2	0	3	0	0	3
CO4	1	3	1	2	2	1	1	0	3	1	2	3
CO5	1	2	2	2	2	2	3	0	3	2	2	3

N	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6						
CO1	<b>(</b> 0)											
CO2	10											
CO3	7											
CO4	2											
CO5												

	Course Code: MAT 2603	Course Title: Number Theory	C	redit	s = 4			
	Course Code: WIA 1 2005	Course Title: Number Theory	L	T	P			
	Elective	Total contact hours: 60	4	0	0			
		List of Prerequisite Courses						
Moder	n Algebra (MAT 2231)							
	List of C	Courses where this course will be prerequisite						
	Description of relevance of	of this course in the M.Sc. Engineering Mathematics Progr	ram		h			
				ر				
		ontents (Topics and subtopics)	1	Hou	ırs			
1	_	thms, Prime and Composite Numbers, Fibonacci and Lucas		8				
	Numbers, Fermat Numbers							
2		GCD, Euclidean Algorithm, Fundamental Theorem of		8				
	Arithmetic, LCM, Linear Dio							
		nodulo n, Linear Congruences, Divisibility Tests, Chinese		12				
3	Remainder Theorem and its applications, Wilson's, Fermat Little and Euler's Theorems							
	with Applications							
4	Multiplicative Functions: F		8					
	Numbers, Möbius Function, I							
5		Order of positive integers, Primality tests, Primitive Roots		8				
6	of Primes, Algebra of Indices	lratic Residues, Legendre Symbols, Quadratic Reciprocity		8				
7		continued Fractions, Infinite continued Fractions		4				
8	Nonlinear Diophantine Equat			4				
0		List of Textbooks/ Reference Books		4				
1		Number Theory with applications, Academic Press, 2 <sup>nd</sup> Ed.						
2		ry Number Theory and Its Applications, Addison Wesley, 5 <sup>th</sup>	Ed					
3		Elementary Number Theory, Springer	Lu.					
4		troduction to the Theory of Numbers, Wiley						
•		rse Outcomes (students will be able to)						
		pepts of divisibility, congruence, greatest common divisor,						
CO1	prime, and prime factorization							
	*	e Reciprocity and other methods to classify numbers as						
CO2	primitive roots, quadratic residues, and quadratic non-residues.							
CO3	- (/ ) -	a to form conjectures about the integers.						
CO4		(proofs) cantered on the material of number theory						
CO5		eory to solve real life problems.						

	<b>&gt;</b>	Mapp	ing of C	ourse O	utcome	s (COs)	with Pro	ogramm	e Outco	mes (POs	)	
Y	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	2	1	0	1	1	1	0	3	1	0	3
CO2	3	2	1	2	0	0	2	0	3	1	0	3
CO3	3	3	2	1	1	0	2	0	3	1	0	3
CO4	2	3	1	2	2	1	1	1	3	1	0	3
CO5	2	2	2	2	2	2	3	2	3	2	1	3

Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)										
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6				

CO1			
CO2			
CO3			
CO4			
CO5			

Approve by Academic Council on Aute.

Approve by Academic Council on Aute.

	Course Codes MAT 2605	Course Titles Crowns and Symmetries	C	redit	s = 4
	Course Code: MAT 2605	Course Title: Groups and Symmetries	L	T	P
	Elective	Total contact hours: 60	4	0	0
		List of Prerequisite Courses			
	List of C	ourses where this course will be prerequisite			
	2.50 07 0	outous more this course marke prevequisite			
	Description of relevance of	f this course in the M.Sc. Engineering Mathematics Progr	am		
					h
	Course Co	ontents (Topics and subtopics)		Hou	rs
1	•	res and polygons. Notions of symmetries in the Euclidean		5	
1	**	Rigid Motions of the Plane. Permutations and bijections			
2	_	groups, Abelian groups, cyclic groups, homomorphisms,		10	)
	kernels, First (and second) iso		10		
3	Group Actions, examples of g		15	j	
4	groups, Symmetric and Dihed		1.5		
4		lane, classification of Euclidean Isometries.		15	
5	Wallpaper Patterns, Frieze pa	v i		10	
6	•	1. C. Escher, Islamic art, African Weavings, Indian Pottery.		5	
7	1 0	oup and symmetries using SageMath			
1		List of Textbooks/ Reference Books			
1	M. A. Armstrong, Groups and David Farmer, Groups and Sy	T T T			
2		roup Theory: An Expedition with SageMath, Narosa			
3	J. A. Gallian, Contemporary				
5	Michael Artin, Algebra, PHI	Austract Argeora, ivarosa			
		se Outcomes (students will be able to)			
CO1		roups and connections with usual notions of symmetry.			
CO2	understand the idea of Group				
CO3	-	x Groups and connections to Linear Algebra			
CO4	understand applications to get				
CO5	Apply SageMath in solving p				
	117 5	<i>U</i> 1 7			
	15/07/07				

		Mapp	ing of C	ourse O	utcomes	s (COs)	with Pro	ogramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	2	1	0	1	1	1	0	3	1	0	3
CO2	3	2	1	2	1	0	2	0	3	1	0	3
CO3	3	3	2	1	1	0	3	0	3	1	0	3
CO4	2	3	2	2	2	1	1	1	3	1	0	3
CO5	2	2	2	2	2	2	3	2	3	2	1	3

N	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6						
CO1												
CO2												
CO3												
CO4												
CO5												

Approve by Academic Council on Aug.

	Comme Code MATERICA	Comment Title Matein Commentations	Hou 8 4 66 8 8 8 6 12		s = 4
	Course Code: MAT 2607	Course Title: Matrix Computations	Ho :	T	P
	Elective	Total contact hours: 60	4	0	0
		List of Prerequisite Courses			
Applie	d Linear Algebra (MAT 2201)				
	List of C	ourses where this course will be prerequisite			
	List of C	ourses where this course will be prerequisite			<b>h</b>
	Description of relevance of	of this course in the M.Sc. Engineering Mathematics Progr	ram	9	
	Description of relevance (	tins course in the Mise. Engineering Mathematics 110g.		3 r	
	Course C	ontents (Topics and subtopics)		Hou	ırs
1		ar transformation and inner product spaces		8	
2	Matrix Norms, Singular Va	lue decomposition, Matrix limit and Markov chain and		4	
2	applications	Ó:		4	
3	Sensitivity of linear Systems,	Sparse matrices and sparse solutions		6	
4	Least Square Problems and va	arious methods to solve		8	
5	Eigenvalue Problems: Unsym	metric and symmetric eigenvalue problems		8	
6		plications, square root of positive semidefinite matrices,		Q	
U	Schur product theorem.			0	
7	Location and Perturbation of			6	
8		on to tensor, rank of tensors, tensor product and		12	
	_	and matricization of tensors with applications			•
	1	List of Textbooks/ Reference Books			
1		Bau, Numerical Linear Algebra, SIAM.			
2		an Loan., Matrix Computations, Johns Hopkins University Pres	SS.		
3		f Matrix Computations, Wiley.			
4	J. Demmel, Applied Numerica	se Outcomes (students will be able to)			
CO1	understand basic concepts in				
CO2	standard matrix norms and its				
CO <sub>2</sub>		real life mathematical problems.			
CO4	understand eigenvalue proble	<u> </u>			
CO <sub>5</sub>		applications to large scale data.			
	understand tensor data and its	applications to large scale data.			

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	2	1	1	1	1	1	1	3	1	0	3
CO2	3	2	1	1	0	0	2	1	3	1	0	3
CO3	3	3	2	1	1	0	2	0	3	1	0	3
CO4	1	3	1	2	2	1	1	1	3	1	1	3
CO5	2	2	2	2	2	2	3	3	3	2	1	3

N	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6						
CO1												
CO2	CO2											

CO3			
CO4			
CO5			

Approve by Academic Council on Aug. Academic Council on Aug.

	G G I MATERIAL	C True C 4	Cr	edits	= 4					
	Course Code: MAT 2621	Course Title: Cryptography	L	P						
	Elective	Total contact hours: 60	4	## T	0					
M - 1	- Alashas (MAT 2221)	List of Prerequisite Courses	1							
Moder	n Algebra (MAT 2231)									
	List of C	Courses where this course will be prerequisite								
	Ziot di C	warses where this course will be prerequisite		-						
	Description of relevance	of this course in the M.Sc. Engineering Mathematics Prog	ram							
	-			)						
	Course C	ontents (Topics and subtopics)		Hour	S					
	Need for cryptography: On ciphertext attacks, Block ciph	line transactions, Perfect secrecy, eavesdropping attacks, ner codes, Hash functions.								
1	Fermat's theorem and Euler	ef introduction to number theory, Euclidean algorithm, Euler's totient func mat's theorem and Euler's generalization, Chinese Remainder Theorem, prim ts and discrete logarithms, Quadratic residues, Legendre and Jacobi symbols.								
2	Private key cryptography:	Stream ciphers, Block ciphers, DES and differential and need encryption standards, Collision resistant hashing,		10						
3	RSA public key cryptosys	tems: RSA system, primality testing, survey of factoring tey cryptosystems: El Gamal public key cryptosystem,		10						
4	Block ciphers, Stream cipher									
5		es: Definition of digital signatures, RSA based digital the Discrete-Logarithm Problem, Certificates and Public-		12						
6	Mathematical Software: Sag	geMath can be used to explore concepts in Cryptography. ged to develop Sage subroutine to solve problems in		15						
		List of Textbooks/ Reference Books								
1	N. Koblitz, A Course in Num	ber Theory and Cryptography, Springer								
2		chot and S. A. Vanstone, Handbook of Applied Cryptography.	, CRC	? Pres	S					
3	4 7 1 2 2 2 1	neory and Practice, CRC Press								
4		luction to Modern Cryptography, CRC Press								
5	Heiko Knopse, A Course in C									
6		ction to Cryptography with Open-Source Software, CRC Pres	SS.							
ad:		rse Outcomes (students will be able to)								
COI	understand various concepts	** * * * *								
CO2	understand various security a									
CO3		otography to real life application.								
CO4	implement Hashing and Digi									
CO5	implement cryptography algo	orithms SageMath and create models.								

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	1	1	1	2	1	1	0	3	1	1	3
CO2	3	1	1	1	0	3	1	1	3	1	0	3

CO3	1	1	1	1	0	1	3	1	3	0	0	3
CO4	1	1	1	3	2	2	1	0	3	0	0	3
CO5	1	3	3	3	2	2	1	1	3	1	1	3

M	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6						
CO1												
CO2												
CO3						0						
CO4						0.2						
CO5						-0						

	G G I MATERIA	C TYOUT I	C	4 0 (					
	Course Code: MAT 2608	Course Title: Topology	L	T	P				
	Elective	Total contact hours: 60	4	0	0				
		<i>Y</i> *							
		List of Prerequisite Courses							
Real an	nd Complex Analysis (MAT 22	230)							
	List of C	Courses where this course will be prerequisite							
	Description of relevance	of this course in the M.Sc. Engineering Mathematics Progr	ram						
			1						
		ontents (Topics and subtopics)		Hou	irs				
1		Sets, Countable and Uncountable Sets, Infinite Sets and		4					
	Axiom of Choice, Well Orde								
2		or a topology, Order topology, Subspace topology, Product		8					
3	topology	Points Continuity Matric Torology and Overlight Torology		1.0					
3	-	Points, Continuity, Metric Topology and Quotient Topology paces, Connected, Subspaces of Real Line, Components and	••						
4	Local Connectedness, simply			8					
		aces, Compact Subspaces of the Real Line, Limit point							
5	compactness, Local Compac			8					
	- , , , , , , , , , , , , , , , , , , ,	ation axioms: Normal Spaces, Urysohn's Lemma (without							
6		orem, Metrization Theorem, Tychonoff's Theorem		8					
_	One-point Compactification	<u> </u>							
7	Characterization of compact	metric spaces, equicontinuity, Ascoli-Arzela Theorem		8					
8	Baire's Category Theorem			4					
<i>&gt;</i>	If time permits, an introducti	on to Fundamental Groups may be covered							
		List of Textbooks/ Reference Books							
1	J. R. Munkres, Topology, 2n	d Edition, Pearson Education (India).							
2	M. A. Armstrong, Basic Top								
3	Stefan Waldman, Topology:	An introduction, Springer.							
4		to Topology and Modern Analysis, McGraw-Hill.							
5		Metric Spaces, 2nd Ed., Narosa Publishing House.							
		rse Outcomes (students will be able to)			· · ·				
CO1	understand different topologi	ical spaces with metric spaces as special cases.							

CO2	identify and learns basic notions of continuity, connectedness, and compactness in	
CO2	arbitrary topological spaces.	
CO3	characterise compact spaces in arbitrary topological spaces.	
CO4	identify Hausdorff, regular and normal spaces.	
CO5	prove an analogy of Bolzano Weirstrass theorem (Arzela Ascolis) theorem for functions	
COS	in the space of continuous functions.	

Mapping of Course Outcomes (COs) with Programme Outcomes (POs)												
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	1	1	1	3	1	1	0	3	0	0	3
CO2	3	2	0	1	2	0	1	0	3	0	0	3
CO3	3	0	1	1	2	0	1	1	3	0	0	3
CO4	3	1	0	1	1	1	1	0	3	2	0	3
CO5	3	2	0	1	2	1	1	0	3	2	0	3

N	<b>Apping of Cou</b>	rse Outcomes (C	COs) with Progr	amme Specific	Outcomes (PSO	(s)
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6
CO1				^		
CO2				. 0		
CO3				•		
CO4				C		
CO5				<i>y</i>		

Course Code: MAT 2609  Elective  Total contact hours: 60  List of Prerequisite Courses  Statistical Computing (MAT 2326), Real and Complex Analysis (MAT 2230)  List of Courses where this course will be prerequisite  NIL  Description of relevance of this course in the M.Sc. Engineering Mathematics Program  This course deals with various real-life application of probability theory in biology, medicine, fin engineering. Several methods taught in Mathematics and Statistics courses in the previous semesters wi in dealing with problems and case studies in this course.	T 0	rs
List of Prerequisite Courses   Statistical Computing (MAT 2326), Real and Complex Analysis (MAT 2230)   List of Courses where this course will be prerequisite   NIL	ance Hour	o and used
Statistical Computing (MAT 2326), Real and Complex Analysis (MAT 2230)  List of Courses where this course will be prerequisite  NIL  Description of relevance of this course in the M.Sc. Engineering Mathematics Program  This course deals with various real-life application of probability theory in biology, medicine, fin engineering. Several methods taught in Mathematics and Statistics courses in the previous semesters wi in dealing with problems and case studies in this course.  Course Contents (Topics and subtopics)  Discrete-Time Markov Models: Discrete-Time Markov Chains, Transient Distributions, Occupancy Times, Limiting Behavior, First-Passage Times.  Poisson Processes, Superposition of Poisson Processes, Thinning of a Poisson Process, Compound Poisson Processes.  Continuous-Time Markov Chains, Transient Analysis: Uniformization, Occupancy Times, Limiting Behavior, First-Passage Times, Birth and Death Processes, Examples of Birth and Death process  Branching Processes, Discrete Time Branching Processes, Generating Function Relations for Branching Processes, Extinction Probabilities  Martingales: super martingales and sub martingales, Optional Sampling theorem, Martingale convergence theorem and their applications  Examples of some stationary processes Mean square prediction of stochastic process, Ergodic theory and stationary processes Mean square prediction of stochastic process, Ergodic theory and stationary processes, properties of Brownian motion, Some  Transformation of Brownian motion, Brownian motion with drift, The Ornstein-Uhlenbeck process  List of Textbooks/ Reference Books  Sheldon M. Ross. Stochastic Processes, 2nd Ed, Wiley.  C. W. Gardiner, Handbook for Stochastic Methods for Physics, Chemistry, and the Natural	Hour	rs
Statistical Computing (MAT 2326), Real and Complex Analysis (MAT 2230)  List of Courses where this course will be prerequisite  NIL  Description of relevance of this course in the M.Sc. Engineering Mathematics Program  This course deals with various real-life application of probability theory in biology, medicine, fin engineering. Several methods taught in Mathematics and Statistics courses in the previous semesters wi in dealing with problems and case studies in this course.  Course Contents (Topics and subtopics)  Discrete-Time Markov Models: Discrete-Time Markov Chains, Transient Distributions, Occupancy Times, Limiting Behavior, First-Passage Times.  Poisson Processes, Superposition of Poisson Processes, Thinning of a Poisson Process, Compound Poisson Processes.  Continuous-Time Markov Chains, Transient Analysis: Uniformization, Occupancy Times, Limiting Behavior, First-Passage Times, Birth and Death Processes, Examples of Birth and Death process  Branching Processes, Discrete Time Branching Processes, Generating Function Relations for Branching Processes, Extinction Probabilities  Martingales: super martingales and sub martingales, Optional Sampling theorem, Martingale convergence theorem and their applications  Examples of some stationary processes Mean square prediction of stochastic process, Ergodic theory and stationary processes Mean square prediction of stochastic process, Ergodic theory and stationary processes, properties of Brownian motion, Some  Transformation of Brownian motion, Brownian motion with drift, The Ornstein-Uhlenbeck process  List of Textbooks/ Reference Books  Sheldon M. Ross. Stochastic Processes, 2nd Ed, Wiley.  C. W. Gardiner, Handbook for Stochastic Methods for Physics, Chemistry, and the Natural	Hour	rs
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Course Contents (Topics and subtopics)	<b>Hou</b> 10	rs
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Uhlenbeck process  List of Textbooks/ Reference Books  Sheldon M. Ross. Stochastic Processes, 2nd Ed, Wiley.  C. W. Gardiner, Handbook for Stochastic Methods for Physics, Chemistry, and the Natural		
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1 Sheldon M. Ross. Stochastic Processes, 2nd Ed, Wiley. C. W. Gardiner, Handbook for Stochastic Methods for Physics, Chemistry, and the Natural		
C. W. Gardiner, Handbook for Stochastic Methods for Physics, Chemistry, and the Natural		
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Third Edition. Springer-Verlag, Berlin.	Scie	nces.
2 V. 1' 1 T. 1 A F'		
3 Karlin and Taylor. A First course in Stochastic Process. Academic Press (Volume-I).		
<ul> <li>Karlin and Taylor. A First course in Stochastic Process. Academic Press (Volume-II).</li> <li>J. Medhi, Stochastic Processes, New Age International.</li> </ul>		
6 Robert P. Dobrow, Introduction to stochastic processes with R-John Wiley & Sons.		
Normal T. J. Bailey, The elements of Stochastic Processes with Application to the Natural	Soio	noos
John Wiley & Sons, Inc.	SCIE	nces.
8 Fima C Klebaner, Introduction to Stochastic Calculus with Applications. 2 <sup>nd</sup> Ed., Imperial Colle	oge F	Press
9 Bernt Oksendal, Stochastic Differential Equations: An Introduction with Applications, Springer		1055.
Course Outcomes (students will be able to)		
CO1 Compute limiting and stationary distribution of Markov chains.		
CO2 Understand the theory and applications of Poisson process.		
CO3 Apply probability generating functions in computations related to branching process.		
CO4 Apply basic inference techniques for making predictions of stochastic process.		
1 11 1		
CO5 Understand the properties of Brownian motion and its application in various real-life		

**Mapping of Course Outcomes (COs) with Programme Outcomes (POs)** 

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	1	3	3	0	0	1	0	3	0	0	3
CO2	0	1	3	3	2	0	1	1	3	0	0	3
CO3	0	0	3	3	1	1	2	2	3	2	0	3
CO4	0	1	3	3	2	2	2	2	3	2	1	3
CO5	0	1	3	3	2	2	2	3	3	3	1	3

Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)										
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6				
CO1						0.2				
CO2										
CO3										
CO4										
CO5						")				

	Course Code: MAT 2630	Course Title: Coding Theory		edits=	1			
			L	T	P			
	Elective	Total contact hours: 60	4	0	0			
		. 0						
		List of Prerequisite Courses						
Basics	linear algebra, and probability	theory.						
		ourses where this course will be prerequisite						
This is	an elective course and not a p	· .						
Description of relevance of this course in the M.Sc. Engineering Mathematics Progr								
This co		ding theory needed for engineering application.						
		Contents (Topics and subtopics)	F	Iours	}			
		nd decoding: Communication channels, Shannon's Theorem,						
1	/	ding, Hamming distance, Nearest neighbor / minimum	ninimum 8					
	distance decoding, Distance of a code							
2	Finite fields and Vector spaces over finite fields							
		Hamming weight, Bases of linear codes, Generator matrix						
3		quivalence of linear codes, encoding with a linear code,		15				
		osets, Nearest neighbor decoding for linear codes, Syndrome		10				
		Dual codes and Reed Muller codes						
		The main coding theory problem, Lower bounds, Sphere-						
4	, ,	Varshamov bound, hamming bound and perfect codes,		11				
A		codes, Plotkin bound, bounds using linear programming,						
	· / ·	codes, Codes and Latin squares.						
<i>&gt;</i>	_	enerator polynomials, Generator and parity check matrices,		10				
5		me special cyclic codes: BCH codes, Definitions, Parameters		10				
		BCH codes, Reed-Solomon codes						
6	Exploration of concepts in co	oding theory using SageMath		10				
		List of Textbooks/ Reference books						
1	J.H. Van Lint, Introduction t	o Coding Theory, Springer						
2	Raymond Hill, A First Cours	se in Coding Theory, Addition-Wesley						
3	San Ling and Chaoping Xing	g, Coding Theory: A First Course, Cambridge University Press	S					

4	Ron M. Roth, Introduction to Coding Theory, Cambridge University Press						
5	Tom Richardson, Rudiger Urbanke, Modern Coding Theory, Cambridge University Press						
6	https://doc.sagemath.org/pdf/en/reference/coding/coding.pdf						
7	https://www.win.tue.nl/~henkvt/images/CODING.pdf						
	Course Outcomes (students will be able to)						
CO1	Use algebraic techniques to construct efficient codes						
CO2	understand vector space over finite fields						
CO3	design linear block codes and cyclic codes						
CO4	understand various error control encoding and decoding techniques						
CO5	develop SageMath codes to solve problems						

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	1	3	3	0	0	1	0	3	0	0	3
CO2	0	1	3	3	2	0	1	0	3	0.	0	3
CO3	0	0	3	3	1	1	2	1	3	2	0	3
CO4	0	0	3	3	2	2	2	2	3	2	1	3
CO5	0	1	3	3	2	2	2	3	3	3	1	3

Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)										
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6				
CO1										
CO2			. 0							
CO3										
CO4										
CO5										

	Comme Code MATIACA	Commer Tilder, A large of Market Alaskar	Cro	edits	= 4				
	Course Code: MAT 2649	Course Title: Advanced Modern Algebra	L	T	P				
	Elective	Total contact hours: 60	4	0	0				
	<b>(</b> 0)								
	List of Prerequisite Courses								
Moder	n Algebra (MAT 2231)								
	A O								
	List of Courses where this course will be prerequisite								
It is a f	It is a foundation course which will be prerequisite for all the course studied in this program								
	Description of relevance of this course in the M.Sc. Engineering Mathematics Program								
,									
	Course C	Contents (Topics and subtopics)	J	Hour	5				
1	Groups: Direct and Semi-dire	ect products of groups, nilpotent and solvable groups.	10						
2	p-groups, Sylow theory, simple groups, structure theorem for abelian groups, introduction to the classification problem for finite groups.								
3	Modules over PIDs, direct sums, simple modules, structure theorem with a focus on vector spaces as modules over polynomial rings.								
4		ory, fundamental theorem, Galois extensions, cyclotomic ns, insolvability of the quintic	20						

	List of Textbooks/ Reference books						
1	J. A. Gallian, Contemporary Abstract Algebra, Narosa						
2	Michael Artin, Algebra, PHI						
3	Dummit and Foote, Abstract Algebra, John Wiley & Sons						
4	G. Santhanam, Algebra, Narosa						
	Course Outcomes (students will be able to)						
CO1	understand the notion of p-groups and Sylow theory.						
CO2	relate semi-direct products to structure theory of groups						
CO3	understand basic results of Module Theory						
CO4	contrast and compare Structure Theorem for Modules over PIDs with the study of structure of linear maps in Linear Algebra.	3					
CO5	develop an understanding of basic Galois Theory and understand its Relation to solving polynomials by radicals.	201					

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	3	2	3	1	0	0	0	3	0	0	3
CO2	3	3	2	3	0	0	1	0	3	0	0	3
CO3	3	3	3	3	0	1	0	1	3	0	0	3
CO4	2	3	3	3	2	3	2	1	3	2	2	3
CO5	2	3	3	3	2	3	2	2	3	2	2	3

M	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)										
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6					
CO1											
CO2		•	O								
CO3		~									
CO4											
CO5											

	Course Code: MAT 2622	Course Title Finite Flowert Method	C	redit	ts = 4			
	Course Code: MAT 2622	<b>Course Title: Finite Element Method</b>	L	T	P			
	Elective	Total contact hours: 60	4	0	0			
		List of Prerequisite Courses						
Differe	ntial Equations (MAT 2235)							
	List of C	ourses where this course will be prerequisite	1		-			
					<u> </u>			
	Description of relevance of	of this course in the M.Sc. Engineering Mathematics Progr	ram	V	,			
1		ontents (Topics and subtopics)		Hou				
1		tional formulation - Rayleigh-Ritz minimization		6				
2	Weighted Residual Approximations: Point collocation, Galerkin and Least Square methods and their applications to the solution of ODE and PDE							
3		Finite Element Formulations for the solutions of ordinary ons: Calculation of element matrices, assembly and solution		1.4	<u>-</u>			
3	of linear equations.		16	)				
	_	sional and two-dimensional basis functions, Lagrange and						
4		s for quadrilaterals and triangular shapes, co-ordinate		16	ń			
_	transformation, integration over a Master Triangular and Rectangular element.							
		tt Method: Finite element solution of Laplace and Poisson						
5		and nonrectangular and curved domains. Applications to		12	2			
		anics and in other engineering problems			-			
	-	o solve some problems on fluid mechanics and in other		(if ti	me			
6	engineering problems using F	•		perm	its)			
		List of Textbooks/ Reference Books	I.					
1	O. C. Zienkiewiez and K. Mo	organ, Finite Elements and approximation, John Wieley.						
2	P.E. Lewis and J.P. Ward, Th	e Finite element method- Principles and applications.						
3	Addison Weley and L. J. Seg	erlind, Applied finite element analysis (2nd Edition), John W	iley					
4	J. N. Reddy, An Introduction	to the Finite Element Method, McGraw Hill, NY.						
5	I.J. Chung, Finite Element Ar	nalysis in Fluid Dynamics, McGraw Hill Inc.						
	Cour	se Outcomes (students will be able to)						
CO1	have basic knowledge in cal-	culus of variation and able to solve ODE and PDE using						
COI	variational methods							
	obtain finite element formulation for ODE using linear and quadratic elements and able							
CO2		Further using given boundary condition, the solution to a						
	given ODE can be obtained.							
		ation for PDE using triangular and rectangular elements						
CO3	_	Il the elements for a given domain. Further, using given						
	-	on to a given PDE can be obtained						
CO4		on from an irregular to a regular domain which will						
	facilitate the computation of i							
CO5	apply the Finite Element Met	hod to some practical problems in 1-D and 2-D problems.						

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	3	0	1	0	2	2	0	3	0	0	3
CO2	0	3	0	1	1	0	3	0	3	0	0	3

CO3	0	3	0	0	0	0	2	0	3	0	0	3
CO4	0	3	0	0	2	3	2	0	3	0	0	3
CO5	0	3	0	0	3	3	3	1	3	3	0	3

	PSO1	PSO2	PSO3	PSO4	PSO5	PSO
CO1						
CO2						
CO3						
CO4						
CO5						-0
APP		caden			And State of the s	

	G G I MATERIA		Cı	redit	s = 4
	Course Code: MAT 2642	Course Title: Integral Transforms	L	T	P
	Elective	Total contact hours: 60	4	0	0
		List of Prerequisite Courses			
Real ar	nd Complex Analysis (MAT 22	230)			
	List of C	Courses where this course will be prerequisite			
					<u> </u>
		of this course in the M.Sc. Engineering Mathematics Prog	_		,
	_	s of various transforms that have immense applications i	n sc	ience	e and
engine	ering, including probability and		<i>P</i>		
		ontents (Topics and subtopics)		Hou	irs
	_	ansforms. Fourier transforms: Introduction, basic properties,		4.0	
1		Ordinary Differential Equations (ODE), Partial Differential		10	,
	Equations (PDE).	i lice			
2	-	tion, differentiation, integration, inverse transform, n's Lemma, solutions to ODE, PDE including Initial Value		1.0	`
2	Problems (IVP) and Boundar		10	,	
	, ,				
3		tion, properties and applications to PDE Mellin transforms: lications; Generalized Mellin transforms. Hilbert transforms		8	
3			0		
		ns; asymptotic expansions of 1-sided Hilbert transforms. n, properties, applications, inversion theorems, properties of			
4		orm. Legendre transforms: Intro, definition, properties,		8	
•	applications	James Begenate transforms. Index, definition, properties,		Ü	
		definition, properties; dynamic linear system and impulse			
5		rms, summation of infinite series, applications to finite		8	
	differential equations				
	Radon transforms: Introd	uction, properties, derivatives, convolution theorem,		0	
6	applications, inverse radon tr	ansform.		8	
7	Wavelet Transform: Discu	ssion on continuous and discrete, Haar, Shannon and		8	
,	Daubechies Wavelets.			0	
	1	List of Textbooks/ Reference Books			
1		opics in Applied Mathematics for Engg. & Physical Scien	ce,	1 <sup>st</sup> ec	lition,
	cambridge:				
2		o Applied Mathematics, Cambridge Press			
3	-	am, Fractional Calculus Theory and Applications of Diffe	erent	iatio	1 and
	Integration to Arbitrary Orde				
4		Handbook of Mathematical Functions, Dover.			
CO1	1	rse Outcomes (students will be able to)			
CO1	=	tial equations using Fourier Transforms.			
CO2		tial equations using Laplace Transforms.			
CO3		ransforms and Hilbert Transforms.			
CO4	solve difference equations us understand wavelet and rado				
CO5	understand wavelet and fadol	i mansivillis.			

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	2	1	0	2	1	1	0	3	0	0	3

CO2	3	2	1	1	3	0	1	0	3	1	0	3
CO3	3	0	1	2	3	0	0	0	3	0	0	3
CO4	3	0	0	0	2	1	1	1	3	0	0	3
CO5	3	0	0	0	2	1	0	2	3	0	0	3

M	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)									
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6				
CO1										
CO2						0				
CO3						-03				
CO4						20				
CO5										

			C	redit	ts = 4					
	Course Code: MAT 2627	Course Title: Mathematical Biology	L	T	P					
	Elective	Total contact hours: 60	4	0	0					
		List of Prerequisite Courses								
Differe	ntial Equations (MAT 2235)									
	List of C	ourses where this course will be prerequisite								
	Description of relevance of	of this course in the M.Sc. Engineering Mathematics Progr	ram							
				Hou						
	Course Contents (Topics and subtopics)									
1	exponential, logistic, Gompertz, ricker growth models, Allee model, Basic dynamical analysis of growth profiles									
2	Harvest models bifurcations and break points discrete time and delay models stable									
3	_	pulations, predator-prey models, host-parasitoid system, of equilibrium points, Poincare-Bendixson's theorem		12	2					
4	Global bifurcations in pre- competition Models	dator-prey models, discrete time predator-prey models,		12	2					
5	structured models and spati models and study of some sta	theory connected to harvest models, An overview of age- ally structured models, concept of stochastic population and and stochastic models in population biology		12	2					
Y		List of Textbooks/ Reference Books								
1		ematical Ecology, Cambridge University Press, Cambridge.								
2		atical Biology, Springer-Verlag, Berlin.								
3		es in Population Biology, Princeton University Press.								
4	Josef Hofbauer, Karl Sigmu Press.	nd, Evolutionary games and population dynamics, Cambrid	lge	Univ	ersity					
5		ological Populations in Space and Time. Cambridge Universit	y Pr	ess.						
6	Stevens, M. Henry, A Primer									
	Cour	rse Outcomes (students will be able to)								

CO1	analyse the mathematical models describing single population dynamics.					
CO2	analyse the mathematical models for interactive population dynamics.					
CO3	understand basic bifurcation theory and apply in population dynamics problems.					
CO4	analyse basic stochastic population dynamics and compute stationary distribution.					
CO5	understand the basic optimal control problem and its application in harvesting models.					
CO6	Construct mathematical models for a given the description of some biological					
C00	phenomena					

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	3	0	0	1	2	2	0	3	0	0	3
CO2	0	3	0	0	1	0	3	0	3	1	0	3
CO3	0	3	0	0	1	0	3	2	3	1	0	3
CO4	0	3	0	0	2	3	2	0	3	0	0	3
CO5	0	3	0	0	3	3	3	1	3	3	0	3
CO6	0	3	0	0	2	3	2	0	3	0	0	3

M	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6						
CO1												
CO2												
CO3												
CO4			70									
CO5												

	Course Code: MAT 2628	Course Title: Signal processing	C	redit	$\frac{ts = 4}{P}$				
	Elective	Total contact hours: 60	4	0	0				
	w3								
Statisti	ical Computing (MAT 2326)								
	:0								
	List of C	ourses where this course will be prerequisite							
	07								
	Description of relevance of	f this course in the M.Sc. Engineering Mathematics Progr	ram						
7									
	Course Co	ontents (Topics and subtopics)		Hot	ırs				
	Review of Linear Continuo	ous-Time Signal Processing: Fourier methods, Laplace							
1	transform, convolution, free continuous filters	equency/time domain processing. Passive and active		8					
_	Sampling and Reconstruction	n: Sampling theorem, aliasing, quantization, sampled data							
2	systems, cardinal (Whitaker) reconstruction, zero, first, second order hold reconstructors,								
2	interpolators, non-resetting	reconstructors, matched filtering. Interpolation and		8					
	decimation.								
3	Discrete-Time Signal Proces	ssing: The z transform, difference equations, relationship		8					

	between F(z) and F*(jw), mappings between s-domain and z-domain, inverse z	
	transform. Discrete–time stability.	
4	Discrete Spectral Analysis: The DFT and its relationship to the continuous FT, the FFT and implementations (decimation in time and frequency), radix-2 implementation, leakage, windowing. Uses of the DFT: convolution — (overlap and add, select savings), correlation. Random processes, power spectral density (PSD) estimation — methods of smoothing the periodogram (Welch's method, windowing the correlation function, etc). ARMA methods.	10
5	Real-Time Simulation Methods Using Difference Equations: Impulse-, step-, ramp-invariant simulations. Tustin's method, matched poles/zeros, bilinear transform methods. Error analysis.	8
6	Filter Design — Continuous and Discrete: Butterworth, elliptic, Chebyshev low-pass filters. Low-pass design methods based on continuous prototypes. Realizations. Conversion to high-pass, band-pass, band-stop filters. Discrete-time filters: IIR and FIR. Linear phase filters. Frequency sampling filters.	10
7	Statistical Signal Processing: Linear prediction, adaptive filters (LMS), recursive least-squares, Nonparametric power spectral density estimation	8
	List of Textbooks/ Reference Books	•
1	Steven B. Damelin, Willard Miller, Jr, The Mathematics of Signal Processing.	
2	Proakis, John G., and Dmitris K. Manolakis. Digital Signal Processing. 4th ed. Upper Sac Prentice Hall.	ldle River, NJ:
3	Oppenheim, Alan V., Ronald W. Schafer, and John R. Buck. Discrete-Time Signal Proce Upper Saddle River, NJ: Prentice Hall	essing. 2nd ed.
	Course Outcomes (students will be able to)	
CO1	Understand the fundamental principles of sampling ideas, Z-transform, discrete frequency related to DSP	
CO2	Understand spectral analysis and estimate the power spectral density by different methods.	
CO3	Understand the designing of filters and test it	
CO4	Understand various real time simulation methods and apply them for real life problems	
CO5	Understand various prediction algorithm for statistical signal processing	

		Mapp	ing of C	course O	utcome	s (COs)	with Pro	ogramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	1	2	1	0	1	0	1	0	3	0	0	3
CO2	0	2	1	2	0	0	2	0	3	0	0	3
CO3	1	3	2	1	1	0	2	1	3	0	0	3
CO4	0	3	1	2	2	1	1	0	3	1	0	3
CO5	0	2	2	2	2	2	3	1	3	2	0	3

M	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)									
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6				
CO1										
CO2						2				
CO3					0					
CO4					6.					
CO5					10					

			C	redit	ts = 4			
	Course Code: MAT 2629	Course Title: Momentum, Heat and Mass Transfer	L	T	P			
	Elective	Total contact hours: 60	4	0	0			
			1 1					
		List of Prerequisite Courses						
Ordina	ary Differential Equation (M	AT 2221), Partial Differential Equations (MAT 2222).						
Nume	rical methods (MAT 2421)	• 0						
	List of (	Courses where this course will be prerequisite						
	<b>Description of relevance</b>	of this course in the M.Sc. Engineering Mathematics Prog	ram					
This c	course deals with several nume	erical and computational techniques of Applied Mathematics	s ha	ving	direct			
implic	implications to industrial and other real life applications.							
	Course Contents (Topics and subtopics)							
1	~	us and curvilinear coordinates		8	1			
	,	wtonian and Non-Newtonian fluids). Deformation, Strain						
2		n tensor, material derivative, steady and unsteady flows,		8				
		on, conservation of mass, potential flows.						
,		I rate of strain, constitutive equation (Newtonian & Non-						
3		' hypothesis, Derivation of Navier-Stokes equation in		12	2			
<i>&gt;</i>		and Spherical Polar system for laminar flows.						
	_	s: Fully developed flow between two parallel plates and						
4		w between two concentric cylinders, flow between two		8				
	concentric rotating cylinders.							
_	1 *	tion of laminar boundary layer equations (using order		0				
5		flow past a semi-infinite flat plate and wedge using		8				
	momentum integral method.	land of back towards and analysis to the state of the sta						
6		law of heat transfer and application to one dimensional and		O				
6	-	Convection of heat. Derivation of equation of energy for		8				
	convective nows in Cartesia	n and cylindrical Polar coordinates, and application to some						

	simple internal flows.							
7	Thermal boundary layer flow past a flat plate and heat transfer in some internal flows	8						
	List of Textbooks/ Reference Books							
1	K. Kundu Pijush, Fluid Mechanics, Elsevier.							
2	G. K. Batchelor, An Introduction to Fluid Dynamics, Cambridge University Press.							
3	H. Schlichting, Klaus Gersten, Boundary-Layer Theory, Springer-Verlag.							
4	S.W. Yuan, Foundations of Fluid Mechanics, Prentice Hall.							
5	R. W. Whorlow, Rheological Technique, Ellis Horwood Ltd.							
6	R.B. Bird, W.E. Stewart E.N., Lightfoot, Transport Phenomena, John Wiley & Sons.							
7	Bennet and Myers, Momentum, Heat and Mass Transfer, Mcgraw Hill, Chemical Engineering Se							
0	1982.							
8	I.G. Currie, Fundamental Mechanics of Fluids, Third edition, 1993,							
	Course Outcomes (students will be able to)							
CO1	develop basic knowledge in tensor analysis and application to various coordinate	)						
	system.							
CO2	develop basic understanding for obtaining governing equation of motion for some							
CO2	specific flow problems.							
CO3	obtain drag coefficient on flow past a rigid body.							
CO4	calculate the heat transfer coefficient and distribution in different materials using heat							
CO4	conduction method.							
CO5	calculate the heat transfer coefficient and distribution in a fluid flow problem.							

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	3	0	0	0	2.	2	0	3	0	0	3
CO2	0	3	0	1	0	0	3	0	3	0	0	3
CO3	0	3	0	0	0	0	3	0	3	0	0	3
CO4	0	3	0	0	2	3	2	0	3	0	0	3
CO5	0	3	0	0.	3	3	3	1	3	3	0	3

N	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)										
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6					
CO1	12										
CO2											
CO3	10										
CO4	7										
CO5											

	Commo Codo, MAT 2050	Course Tides Democrated on Theory	C	Credits = 4				
	Course Code: MAT 2650	Course Title: Representation Theory	L	T	P			
	Elective	Total contact hours: 60	4	0	0			
		List of Prerequisite Courses						
Moder	rn Algebra (MAT 2231)							
	List of Courses where this course will be prerequisite							

	Description of relevance of this course in the M.Sc. Engineering Mathematics Progr	ram
	Course Contents (Topics and subtopics)	Hours
1	Review of Group Actions. Groups acting on vector spaces (Matrix Groups). General Linear group and its subgroups.	5
2	Representations of a group, finite dimensional representations, one-dimensional representations. New representations from old, direct sums, tensor products, sub-representations.	10
3	Maschke's Theorem, Schur's Lemma, Irreducible representations, Complete reducibility.	15
4	Matrix elements, Characters of a representation, Orthogonality relations, regular representations, counting irreducible representations.	20
5	Representations of the symmetric group, and applications, Computation of Young Tableaux.	10
	List of Textbooks/ Reference Books	
1	G. James and M. Liebeck, Representations of Finite Groups, Cambridge University Press.	
2	J. P. Serre, Linear Representations of Finite Groups, GTM Springer	
3	C. S. Musili, Representations of Finite Groups, TRIM Series	
4	Alperin and Bell, Groups and Representations, GTM Springer	
5	Dummit and Foote, Abstract Algebra, John Wiley & Sons	
6	M. Artin, Algebra, PHI	
	Course Outcomes (students will be able to)	
CO1	understand the basic notions and constructions of representations.	
CO2	understand the role played by character theory	
CO3	understand the representation theory of Abelian groups.	
CO4	understand the basic ideas in the representation theory of symmetric groups.	
CO5	understand some simple applications of representation theory.	

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)												
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	
CO1	0	3	0	4	0	2	2	0	3	0	0	3	
CO2	0	3	0	Q1	0	0	3	0	3	0	0	3	
CO3	0	3	0	0	0	0	3	0	3	0	0	3	
CO4	0	3	0	0	2	3	2	0	3	0	0	3	
CO5	0	3	0	0	3	3	3	1	3	3	0	3	

M	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
201	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6						
CO1												
CO2												
CO3												
CO4												
CO5												

	Course Code: MAT 2610	Course Title: Advanced Mathematical Finance	C	redi	ts = 4
	Course Code: MAI 2010	Course Title: Advanced Mathematical Finance	L T	P	
	Elective	Total contact hours: 60	4	0	0

#### **List of Prerequisite Courses** Financial Mathematics (MAT 2606), Statistical Computing (MAT 2326), Stochastic Process (MAT 2609) List of Courses where this course will be prerequisite Description of relevance of this course in the M.Sc. Engineering Mathematics Program This course gives students an exposure to applications of mathematics in baking and finance section. Students get the exposure to stochastic differential equation, Ito calculus and pricing of various financial instruments. **Course Contents (Topics and subtopics)** Hours Review of Probability Spaces and Convergence concepts, Filtrations, Expectations, 1 8 Change of Measures Brownian motion calculus, Ito Integral and its properties, Ito processes and Stochastic 2 12 differentials, Ito formula for Ito processes and Martingale properties. Stochastic Differential Equations, existence, and uniqueness, Backward and Forward 3 equations, numerical techniques for simulation of stochastic differential equations, 12 Multivariate extensions Risk neutral pricing in discrete time and continuous time, Stock and FX options, financial derivatives and arbitrage, Semi martingale market model, Diffusion and Black 10 4 Scholes model and other examples Applications to Bonds, Rates and Options, Bonds and Yield curve, Models based on spot 5 10 rates, Merton's model and Vasicek's model 6 Numerical Schemes for simulation of Stochastic differential equations 8 Software: R/Python List of Textbooks/ Reference Books Fima C Klebaner, Introduction to Stochastic Calculus with Applications, Second edition, Imperial College Press. Steven Shreve, Stochastic Calculus for Finance I: The Binomial Asset Pricing Model, Springer. 2 3 Steven Shreve, Stochastic Calculus for Finance Continuous-Time Models, Springer. Fima C Klebaner, Introduction to Stochastic Calculus with Applications. Second Edition, Imperial 4 College Press. Peter E. Kloeden, Eckhard Platen, Henri Schurz, Numerical Solution of SDE Through Computer 5 Stefano M. Iacus, Simulation and Inference for Stochastic Differential Equations with R Examples, 6 Zdzisław Brzeźniak and Tomasz Zastawniak, Basic Stochastic Processes: A Course Through Exercises, Springer. **Course Outcomes (students will be able to...)** understand basic theory of Ito processes and Ito integrals. CO1 solve basic stochastic differential equations and properties of solutions. CO2 CO3 simulate numerical solutions of some simple stochastic differential equations. CO4 apply Ito stochastic calculus for pricing financial instruments.

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	2	1	1	1	0	1	0	2	0	0	3
CO2	0	2	1	2	0	0	2	0	2	0	0	3
CO3	1	3	2	1	1	0	2	0	2	0	0	3

CO5

apply the methods to analyse real data sets from financial markets.

CO4	0	3	1	2	1	10	1	0	2	1	2	3
CO5	0	2	2	2	1	2	3	1	2	2	2	3

M	<b>Sapping of Cou</b>	rse Outcomes (C	COs) with Progr	amme Specific (	Outcomes (PSO	s)
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6
CO1						
CO2						
CO3						
CO4						0-
CO5						0.2

			\		
	Course Code: MAT 2625	Course Title Multivariete Analysis	C	redit	s = 4
	Course Code: MA1 2025	Course Title: Multivariate Analysis	L	T	P
	Elective	Total contact hours: 60	4	0	0
		List of Prerequisite Courses			
Statist	ical Computing (MAT 2326), P	Programming Lab (MAP 2521)			
		, 0'			
	List of C	Courses where this course will be prerequisite			
		20,			
	Description of relevance of	of this course in the M.Sc. Engineering Mathematics Progr	ram		
With	an enormous increase of the	large-scale computational methods in science and engine	erin	g, aj	pplied
mathe	maticians must get exposure to	various statistical methods. This course aims to give the students	dent	s exp	osure

to the theory of multivariate statistics and their applications in real life problems.

	Course Contents (Topics and subtopics)	Hours
1	Review of linear algebra, review of multivariate distributions, multivariate normal distribution and its properties, distributions of linear and quadratic forms	8
2	Tests for partial and multiple correlation coefficients and regression coefficients and their associated confidence regions. Data analytic illustrations	8
3	Wishart distribution (definition, properties).	6
4	Construction of tests, union-intersection and likelihood ratio principles, inference on mean vector, Hotelling's T <sup>2</sup> , MANOVA	8
5	Inference on covariance matrices. Discriminant analysis. Principal component analysis and factor analysis	10
6	Multivariate Linear Regression, Practical on the above topics using statistical packages for data analytic illustrations,	10
7	Clustering, Distance methods and Ordination and application to real data sets.	10
	List of Textbooks/ Reference Books	
1	T. W. Anderson, An Introduction to Multivariate Statistical Analysis.	
2	R. A. Johnson and D. W. Wichern, Applied Multivariate Statistical Analysis.	
3	K. V. Mardia, J. T. Kent and J. M. Bibby, Multivariate Analysis.	
4	M. S. Srivastava and C. G. Khatri, An Introduction to Multivariate Statistics.	
	Course Outcomes (students will be able to)	
CO1	Illustrate the geometry of sample and various properties of multivariate normal distribution	
CO2	Apply various testing procedures for multivariate data	
CO3	Derive the sampling distribution of statistics and apply them to construct testing procedures in a multivariate set up	

CO4	Understand and apply multivariate regression methods to solve real life problems	
CO5	Apply various multivariate methods using statistical packages to solve real life problems	
CO6	Understand and apply various clustering method in multivariate data sets.	

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	0	3	3	2	0	2	0	3	0	0	3
CO2	0	0	3	3	2	2	2	0	3	1	0	3
CO3	0	0	3	3	2	0	1	0	3	0	0	3
CO4	0	0	3	3	3	1	3	3	3	0	2	3
CO5	0	0	3	3	2	3	3	3	3	2	2	3
CO6	0	0	3	3	2	1	3	3	3	0	2	3

N	<b>Tapping of Cou</b>	rse Outcomes (C	COs) with Prog	ramme Specific	Outcomes (PSO	S)
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6
CO1				-	7	
CO2				\	,	
CO3				-0		
CO4				, 0'		
CO5						
CO6				70,		

	Comme Code MAT 202	Commo Title Design and Ameloric of Francis	C	redit	s = 4
	Course Code: MAT 2626	Course Title: Design and Analysis of Experiments	L	T	P
	Elective	Total contact hours: 60	4	0	0
		20'			
		List of Prerequisite Courses			
Applie	d Linear Algebra (MAT 2201).	Statistical Computing (MAT 2326)			
	List of C	ourses where this course will be prerequisite			
	107				
	Description of relevance of	of this course in the M.Sc. Engineering Mathematics Progr	ram		
	Course C	Contents (Topics and subtopics)		Hou	ırs
		÷			
1. A	classification model. Analys	rkoff Theorem, Randomization and Replication, Analysis of one-way on model. Analysis of two-way classification model with equal number of		16	5
	observations per cell with a	nd without interactions. Analysis of two-way classification		0	,
Y	-	of observations per cell without interactions			
2	Analysis of BIBD. Analysi	s of covariance in one way and two-way classification		10	)
		es for estimable parametric functions.		- 10	,
3	1	s, 2Kdesign, confounding in 2K design, Partial confounding		10	)
	and total confounding			10	,
	Response surface methodolo	ogy (RSM): linear and quadratic model, stationary point,			
4		CD), ridge systems, multiple responses, concept of rotatable		16	í
7	designs, Box-Behnken desig	en, optimality of designs, simplex lattice designs, simplex		1(	,
	centroid designs				

5	Taguchi methods: concept of noise factors, concept of loss function, S/N ratio,	8
3	orthogonal arrays	o
6	Software: R/Python/MATLAB	
	List of Textbooks/ Reference Books	
1	Montgomery, D.C. Design and Analysis of Experiments, Wiley.	
2	Dean, A. and Voss, D. Design and Analysis of Experiments, Springer	
3	George E. P. Box, Draper N.R. Empirical Model-Building and Response Surfaces, Wiley	
4	W. W. Hines, D. C. Montgomery, Probability and Statistics in Engineering. John Wiley.	
5	Rao, C. R. Linear Statistical Inference and Its Applications, Wiley	
	Course Outcomes (students will be able to)	0
CO1	perform statistical analysis of one-way and two-way classified data.	
CO2	analyse data coming from factorial experiments.	-O
CO3	understand basic principles of response surface methodology and apply them in real	
CO3	life problems.	
CO4	apply Taguchi methods to optimize designs.	,
CO5	use statistical software to analyse real data and interpret the results.	

		Mapp	ing of C	ourse O	utcomes	s (COs)	with Pro	ogramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	0	3	3	0	1	1	0	3	0	0	3
CO2	0	0	3	3	2	3	3	3	3	2	0	3
CO3	0	0	3	3	2	3	3	3	3	2	0	3
CO4	0	0	3	3	1	3	3	3	3	2	2	3
CO5	0	0	3	3	1	3	3	3	3	2	2	3

		•	C								
N	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)										
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6					
CO1											
CO2		~~									
CO3	7										
CO4	4	Y .									
CO5	12										

	Correct Code MAT 2022	Comment Trials Comment on Description	Cr	edits	s = <b>4</b>
	Course Code: MAT 2623	Course Title: Operation Research	L	T	P
	Elective	Total contact hours: 60	4	0	0
		List of Prerequisite Courses			
Applie	d Linear Algebra (MAT 2201)	, Optimization techniques (MAT 2232)			
	I :- 4 - 6 C	N			
	List of C	Courses where this course will be prerequisite			
	Decemintion of volevence	of this course in the M.Sc. Engineering Mathematics Progr	10.771		
	Description of relevance	of this course in the M.Sc. Engineering Mathematics 1 rogi	am		)
	Course C	Contents (Topics and subtopics)		Hou	rs
1		action of operation research using historical perspective		4	
		m: Simplex Methods, revised simplex method, two phase			
2		hod, Karmakar Method, Sensitivity analysis and Duality		12	
3	Integer Programming	nod, Harmakar Mediod, Benoitivity unaryons and Balancy		8	
		Characteristics of dynamic programming, Dynamic			
4		Priority Management employment smoothening, capital		8	
		rtest Path, cargo loading and Reliability problems			
	Transportation and Assignm	nent Problems: Transportation Problems definition, Linear			
	form, Solution methods: 1	North-west corner method, least cost method, Vogel's			
5	approximation method. Deg	generacy in transportation, Modified Distribution method,		12	
	Unbalanced problems and	profit maximization problems. Transhipment Problems			
		ravelling sales man problems.			
6	-	ry classification, Different cost associated to Inventory,		4	
Ü		ventory models with deterministic demands, ABC analysis.			
		Queuing theory, elements of queuing theory, Kendall's			
7		teristics of a queuing system, Classification of Queuing		8	
	models and preliminary exam	nples.			
8	Network models	List of Textbooks/ Reference Books		4	
1	Hamilton Danieliana D				
1		search: An Introduction, Pearson.	· · ·		
2		mani, A Tamilarasi, Operations Research, Pearson Education I		1+h	Ed
3	Cengage Learning.	. Venkataramanan, Introduction to Mathematical Programm	mng,	, 4ui	Eu,
4		l-Louis, Operations Research-A Model Based Approach, Sprin	nger		
		na Dev Kumar, Introductory Operations Research, Theory and		licat	ions
5	Springer.	a 20. Italia, intoductory operations research, theory and	• • • PP	. 110al	10110,
		rse Outcomes (students will be able to)			
CO1		the subject of operation research.			
CO2	•	oblems arising in science and engineering.			
CO3	1 0 01	solve linear programming problems.			
CO4		as linear programming or dynamic programming problems.			
CO5		problems arising in science and engineering.			

		Mapp	ing of C	ourse O	utcomes	s (COs)	with Pro	ogramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	1	3	3	0	1	1	0	3	0	0	3
CO2	0	0	3	3	1	3	3	3	3	2	0	3

CO3	0	1	3	3	2	3	3	3	3	2	0	3
CO4	0	2	3	3	1	3	3	3	3	2	2	3
CO5	0	3	3	3	1	3	5	5	3	2	2	3

CO1	CO2 CO3 CO4		PSO1	PSO2	PSO3	PSO4	PSO5	PSO
CO3	CO3	CO1						
CO4 CO5 CO5 CO4 CO5 CO5 CO4 CO5	CO4 CO5 CO5 CO4 CO5 CO4 CO5 CO4 CO5 CO4 CO5 CO4 CO5 CO4 CO5 CO5 CO4 CO5 CO5 CO4 CO5 CO5 CO4 CO5							
Cos Council on Auto-	COS COUNCIL ON AUGO	CO4						0
Academic Council on Aug.	orove by Academic Council on Aug.	CO5						
	OTO TE					icil on	Tigo.	
		P.O.O.Y	Je of	Caden				
			3603	eaden?				
		APP	203	Caden				
		A POT	204	caden				

	Course Code: MAT 2644	Course Titles Coometwy of Courses and Southern	C	redit	s = 4
	Course Code: MAT 2644	Course Title: Geometry of Curves and Surfaces	L	T	P
	Elective	Total contact hours: 60	# Hou 8 8 12 10 10 K Peters, L tion	0	
		List of Prerequisite Courses			
Real a	nd Complex Analysis (MAT 22	02)			
	List of C	ourses where this course will be prerequisite	1		
	Description of relevance of	of this course in the M.Sc. Engineering Mathematics Progr	ram		<b>_</b>
	<u> </u>	4 (D) 1 1 1 1 1 1			
		ontents (Topics and subtopics)	1	Hou	irs
1	formulas, Fundamental Theor	pace curves: Curvature and torsion of curves, Serret-Frenet em of space curves.		8	
	1	Change of parameters, Differentiable functions, Tangent			
2	plane, Differential of a map s	urfaces, Orientable surfaces		8	
	First and second fundamen	tal Form: The first fundamental Forms, The Gauss map,			
3	The second fundamental form	ns, Normal and principal curvatures, introduction to ruled		12	ļ
	and minimal surfaces				
4	Curves on Surfaces: Curvatu			12	
	_	s of Surfaces: Tensor Notation, Gauss's Equations and the			
5	1	Equations and the Theorema Egregium, The Fundamental		10	)
	Theorem of Surface Theory				
6	1	s applications to surfaces of constant curvatures		10	<u> </u>
	•	List of Textbooks/ Reference Books			
1	_	n Lovett, Differential Geometry of Curves and Surfaces, A K		ers, I	∠td.
2	-	ves and Surfaces, by Manfredo P. Do Carmo, Dover Publicat	ion		
3		Geometry of Curves and Surfaces, Springer			
5	_	fferential Geometry, Cambridge University Press  Differential Geometry, Springer.			
3		st Course in Curves and Surfaces, by Theodore Shifrin, whi	oh i	0.0110	ilabla
6	free online at <a href="http://math474.">http://math474.</a>		CII I	s ava	паоте
		se Outcomes (students will be able to)			
CO1	V .	theory of plane and space curves.			
CO2	understand theory of surfaces	• 1			
CO3		vature of curves and surfaces.			
CO4	- 1/1	ompute curvatures of curves and surfaces.			
CO5	analyse curves and surfaces a	<u> </u>			
			<u> </u>		

_ <	<b>)</b> Y	Mapp	ing of C	ourse O	utcome	s (COs)	with Pro	ogramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	3	0	0	2	0	2	0	3	0	0	3
CO2	3	3	0	2	2	0	1	0	3	0	0	3
CO3	3	3	1	2	2	0	2	0	3	1	0	3
CO4	3	3	0	0	2	2	3	0	3	0	0	3
CO5	3	3	0	0	2	2	3	0	3	1	0	3

N	<b>Sapping of Cour</b>	rse Outcomes (C	COs) with Progra	amme Specific (	Outcomes (PSO	es)
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6

CO1			
CO2			
CO3			
CO4			
CO5			

Approve by Academic Conneil on Aute.

Approve by Academic Conneil on Aute.

	Course Code: MAT 2645	Course Title: Convex Optimization	Cro	edit	s = 4
		<u> </u>	L '	Т	P
	Elective	Total contact hours: 60	4	0	0
		List of Prerequisite Courses	I		
Applie	d Linear Algebra (MAT 2201),	Optimization Techniques (MAT 2232)			
	List of C	langaa whana thia aangaa will ba muanaaniaita			
	List of C	ourses where this course will be prerequisite			
	Description of relevance of	of this course in the M.Sc. Engineering Mathematics Progr	ram		
			(	7	7
	Course C	ontents (Topics and subtopics)		Hou	rs
1	Introduction to Convex optim	nization problems		4	
2	Convex sets: Affine and	convex sets with examples, operations that preserves		10	
2	convexity, generalized inequa	ality, separating and supporting cones, dual cones		10	<u> </u>
		on and examples of convex functions, operations that			
3		ate and quasi conjugate functions, log concave and convex		8	
	function				
		mization problems: Generalized optimization and convex			
4		examples. Linear and quadratic optimization problems,		10	)
	Geometric programming prob	and geometric interpretation, Optimality conditions,			
5	perturbation and sensitivity a			10	)
	-	mization: Approximation and fitting, Statistical estimation,			
6	Geometric problems	, see		10	1
	Interior point methods: Ine	quality constrained minimization problems, Logarithmic			
7	barrier function and central j	path, The barrier method, Feasibility and phase I methods,		12	ļ
	_	equalities, Primal-dual interior-point methods			
	Mathematical software: Pytho				
	1	List of Textbooks/ Reference Books			
1		ndenberghe, Convex Optimization, Cambridge University Pre	ess		
2	R. T. Rockafellar, Convex A				
3		di Nemirovski, Lectures on Modern Convex Optimizat	ion:	Ana	ılysis,
4		Applications, SIAM Publication			
4		n Lewis, Convex Analysis and Nonlinear Optimization, Springse Outcomes (students will be able to)	ger		
CO1	understand basic convex opti				
CO2		convex optimization problems.			
CO <sub>2</sub>	1	oblems using standard algorithms.			
CO4		hods to solve convex optimization problems.			
CO5	· · · · · · · · · · · · · · · · · · ·	ization to solve real world problems.			

		Mapp	ing of C	Course O	utcome	s (COs)	with Pro	ogramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	3	3	3	3	0	2	0	3	0	0	3
CO2	0	3	3	3	2	0	2	0	3	0	0	3
CO3	1	2	3	3	2	0	2	0	3	0	0	3
CO4	0	2	3	3	3	0	2	0	3	0	0	3
CO5	0	3	3	3	3	3	3	3	3	3	1	3

N	<b>Iapping of Cour</b>	rse Outcomes (C	COs) with Progr	amme Specific	Outcomes (PSC	Os)
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6
CO1						
CO2						2
CO3					0	
CO4					6.	
CO5					10	

	Course Code: MAT 2646	Course Title: Time Series Analysis	Cr	edit	s = 4		
	Course Code: MAI 2040	Course Title: Time Series Analysis	L	T	P		
	Elective	Total contact hours: 60	4	0	0		
		List of Prerequisite Courses					
Statisti	ical Computing (MAT 2326)						
		.0					
	List of C	ourses where this course will be prerequisite					
		10 Y					
		of this course in the M.Sc. Engineering Mathematics Prog					
	7	ply various time series models for forecasting problems which	ch ab	unda	ant in		
industr							
		ontents (Topics and subtopics)	]	Hou	rs		
1	-	e series: Graphical display, classical decomposition model,	4				
	1 2	and cycle, estimation of trend and seasonal components.	<u> </u>				
	Stationary time series models: Concepts of weak and strong stationarity, AR, MA and						
2	1	properties, conditions for stationarity and invertibility,		12			
		CF), partial autocorrelation function (PACF), identification					
		timation, order selection and diagnostic tests.					
	7	ry models: ARIMA model, determination of the order of					
3		and difference stationary processes, tests of nonstationarity		12			
	_	Fuller (DF) test, augmented DF test, and Phillips-Perron test					
4		ential smoothing, Holt-Winters method, minimum MSE		10			
	1	mple and out-of-sample forecast.					
5		es: Seasonal ARIMA models, estimation; seasonal unit root		6			
	test (HEGY test).						
6		: State space representation of ARIMA models, basic		8			
-	structural model, and Kalmar	<u> </u>					
7	Spectral analysis of weakly	8					
•	its properties, s. d. f. of AF	R, MA and ARMA processes, Fourier transformation and					

	periodogram
8	Statistical software: R/Python
	List of Textbooks/ Reference Books
1	P. Brockwell and R. Davis, Introduction to Time Series and Forecasting, Springer, Berlin.
2	Box, G. Jenkins and G. Reinsel, Time Series Analysis-Forecasting and Control, 3rd ed., Pearson
2	Education.
3	W. A. Fuller, Introduction to Statistical Time Series.
4	Ruey S. Tsay, An Introduction to Analysis of Financial Data with R, John Wiley.
5	T. W. Anderson, The Statistical Analysis of Time Series.
6	R. H. Shumway and D. S. Stoffer, Time Series Analysis and Its Applications.
7	C. Chatfield, The Analysis of Time Series – An Introduction, Chapman and Hall / CRC, 4th ed
	Course Outcomes (students will be able to)
CO1	apply graphical techniques to descriptive exploration of time series data.
CO2	understand different statistical properties of stationary time series models and apply
CO2	them in analysing real data.
CO3	apply different forecasting techniques for time series data.
CO4	apply state space models in forecasting problems.
CO5	compute spectral density functions for different time series models.

		Mapp	ing of C	course O	utcome	s (COs)	with Pro	gramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	0	3	3	1	0	2	0	3	0	0	3
CO2	0	0	3	3	2	2	2	0	3	1	0	3
CO3	0	0	3	3	2	1	2	0	3	1	0	3
CO4	0	0	3	3	2	1	3	3	3	0	1	3
CO5	0	0	3	3	1	$\bigcup 0$	3	3	3	2	1	3

N	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6						
CO1	7											
CO2	4	¥ *										
CO3	1											
CO4												
CO5	10											

	Course Code: MAT 2611		C	redi	ts = 4
K,		<b>Course Title: Computational Fluid Dynamics</b>	L	T	P
	Elective	Total contact hours: 60	4	0	0
			•	·	
		List of Prerequisite Courses			
Differe	ential Equations (MAT 2235), A	Advanced Differential Equations (MAT 2233)			
	List of C	ourses where this course will be prerequisite			
	Description of relevance of	f this course in the M.Sc. Engineering Mathematics Pro	gram		
This co	ourse deals with several nume	rical and computational techniques of Applied Mathemat	ics ha	ving	direct

implica	ations to industrial and other real-life applications.	
	Course Contents (Topics and subtopics)	Hours
1	Introduction to tensor calculus and curvilinear coordinates	8
2	Classification of fluids (Newtonian and Non-Newtonian fluids). Deformation, Strain	
	tensor, Rate of deformation tensor, material derivative, steady and unsteady flows,	6
	streamline and stream function, conservation of mass, potential flows.	
3	Relation between stress and rate of strain, constitutive equation (Newtonian & Non-	
	Newtonian fluids), Stokes' hypothesis, Derivation of Navier-Stokes equation in	10
	Cartesian, Cylindrical Polar and Spherical Polar system for laminar flows.	
4	Flow in some simple cases: Fully developed flow between two parallel plates and	
	through circular pipe, Flow between two concentric cylinders, flow between two	6
	concentric rotating cylinders.	CV.
5	Grid generation, Structured and Unstructured grid generation methods	6
6	Solution of Systems of Linear Algebraic Equations using iterative methods such as:	
Ü	Gauss-Seidel iterative method, Line by line TDMA, ADI (Alternating direction implicit)	
	method etc. Stability and convergence of numerical methods. Finite Volume	10
	Discretization of 1-D, 2-D and 3-D problems. Application of various iterative methods	
	to the discretized Equations.	
7	Finite volume discretization of convection-diffusion problem: Central difference	
•	scheme, Upwind scheme, Power-law scheme, Generalized convection-diffusion	4
	formulation.	
8	Finite volume discretization of two-dimensional convection-diffusion problem, the	
Ü	concept of false diffusion, Discretization of the Momentum Equation: Stream Function	
	Vorticity approach and Primitive variable approach, Staggered grid, SIMPLE,	10
	SIMPLER algorithm etc.	
	List of Textbooks/ Reference Books	
1	Pijush K. Kundu and Ira M Cohen, Fluid Mechanics, Elsevier.	
2	G. K. Batchelor, An Introduction to Fluid Dynamics, Cambridge University Press.	
3	S.W. Yuan, Foundations of Fluid Mechanics, Prentice Hall.	
4	R. W. Whorlow, Rheological Technique, Ellis Horwood Ltd.	
5	R.B. Bird, W.E. Stewart E.N., Lightfoot, Transport Phenomena, John Wiley & Sons.	
6	Fletcher C.A.J, Computational Techniques for Fluid Dynamics, Volumes I & II, Springer-	Verlag
7	C. Hirsch, Numerical Computation of Internal and External Flows, Volume I & II, Wiley.	veriug.
8	J. C. Tannehill, D. A. Anderson and R. H. Pletcher, Computational Fluid Mechanics and	Heat Transfer
O	McGraw-Hill.	Ticat Transici
9	G. D. Smith, Numerical Solution of Partial Differential Equations: Finite Difference	Mathada Nas
9	York, NY: Clarendon Press.	wichiods, ive
10	M. Schafer-Computational engineering- Introduction to numerical methods.	
11	M. Farrashkhalvat, J Miles, Basic Structured Grid Generation, Elsevier.	
12	S. V. Patankar, Numerical Heat Transfer and Fluid Flow, Hemisphere Pub.	
		Crow 11:11
13	John. D. Anderson, Jr., Computational Fluid Dynamics, The Basics with Applications, Mc	Graw-Hill.
<i>&gt;</i>	Course Outcomes (students will be able to)	
CO1	develop basic knowledge in tensor analysis and application to various coordinate	
	system	
CO2	develop basic understanding for obtaining governing equation of motion for some	
	specific flow problems. And obtain velocity profiles and drag coefficient.	
965	generate the grids in different coordinate system and apply various iterative methods	
CO3	to a large system of linear and non-linear algebraic equations, which will guarantee	
	the convergence of the system.	
CO4	discretise ODE and PDE using finite volume method and will be able to solve the	
	discretised linear equation using various iterative methods along with boundary	

	conditions.	
CO5	apply finite volume method to discretise laminar fluid flow problems using upwind, hybrid and power-law schemes along with SIMPLE and SIMPLER algorithms and use of various programming languages such as: PYTHON, MAT LAB, FLUENT etc. to obtain the numerical solutions to the discretised	

		Mapp	ing of C	ourse O	utcome	s (COs)	with Pro	ogramm	e Outco	mes (POs	)	
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	0	3	0	0	0	2	3	1	3	0	0	3
CO2	0	3	0	0	0	0	3	1	3	0	0	3
CO3	0	3	0	0	0	0	3	0	3	0	0	3
CO4	0	3	1	0	2	3	2	0	3	1	1	3
CO5	0	3	0	0	3	3	3	1	3	3	0	3

N	Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)											
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6						
CO1				7								
CO2				7	,							
CO3												
CO4				.1								
CO5												

	Common Codo, MAT 2047	Course Title: On oreter Theory	C	redit	ts = 4
	Course Code: MAT 2647	Course Title: Operator Theory	L	T	P
	Elective	Total contact hours: 60	4	0	0
		List of Prerequisite Courses			
Applie	d Linear Algebra (MAT 2201)	Real and Complex Analysis (MAT 2230)			
	List of C	Courses where this course will be prerequisite			
Not Ap	pplicable				
	<b>Description of relevance</b>	of this course in the M.Sc. Engineering Mathematics Progr	ram		
	10				
	Course C	Contents (Topics and subtopics)		Hou	ırs
1		rt spaces, Dual spaces and transposes, Orthonormal basis.		15	5
A	Projection theorem and Riesz	*			
2	Hyponormal operators.	ators on a Hilbert space, Normal, self-adjoint unitary,		10	)
3	Spectrum of bounded operator	ors and numerical ranges		10	)
4	Theory of Compact operators	s on normed spaces and its spectrum.		1(	)
5	Spectral theorem for compac	t self-adjoint operators and Singular value decomposition		15	;
		List of Textbooks/ Reference Books			
1	B.V. Limaye, Functional Ana	alysis, 2nd Edition, New Age International.			
2	J. B. Conway, A Course in F	unctional Analysis, 2 <sup>nd</sup> Edition, Springer.			
3	_	f Operator Theory, Birkhauser.			
4		ctional Analysis with Applications, John Wiley & Sons.			
5	S. G. Mikhlin, Variation Met	thods in Mathematical Physics, Pergaman Press, Oxford.			

	Course Outcomes (students will be able to)					
CO1 identify various operators on Hilbert spaces.						
CO2	compute spectrum of operators.					
CO3	understand the spectral theorem of compact operators and apply it to prove the singular					
CO3	value decomposition.					
CO4	apply the theory to Sturm Liouville boundary value problems.					
CO5	see the analogy between polar representation of complex numbers and operators.					

	Mapping of Course Outcomes (COs) with Programme Outcomes (POs)											
	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	3	2	2	0	2	0	0	0	3	0	0	3
CO2	3	2	0	0	3	0	2	0	3	1	0	3
CO3	3	0	2	1	3	0	0	0	3	0	0	3
CO4	3	0	0	0	2	1	1	0	3	0	0	3
CO5	3	0	0	0	2	2	0	0	3	00°	0	3

Mapping of Course Outcomes (COs) with Programme Specific Outcomes (PSOs)						
	PSO1	PSO2	PSO3	PSO4	PSO5	PSO6
CO1				. 0		
CO2						
CO3				C		
CO4				<b>Y</b>		
CO5						

## Marks distribution for On Job Training (OJT) (MAP 2811)

- At the end of OJT, students will have to submit (i) a written report of the work carried out, and (ii) Evaluation of the student from the Industry Mentor. After coming back to the Institute, the student would have to present the work carried out to a committee of two faculty members of the Institute. The presentation would be evaluated by the committee and students will be given a grade for the OJT based on the following parameters.
- Format of the evaluation by the industry mentor:

Name of the Student	
Name and designation of the	
mentor	
Name and address of the	
organization/ place of internship	
Email of the mentor	
Phone number	
Internship start date	
Internship end date	

• Instruction to Mentor: Please evaluate the student on following Parameters & tick appropriate column:

Excellent: > 80%, Good: 60 - 80%, Satisfactory: 40 - 60%, Needs Improvement: < 40%

	Needs improvement	Satisfactory	Good	Excellent
General behaviour:				

ethics and		
attendance		
Oral and written		
communication		
skills		
Technical		
knowledge		
Interpersonal skills		-3
Professional skills:		
Initiative and		
motivation		, V
Managerial skills:		
Time and resource		

Any other remarks:

Signature of the mentor with date:

• Format for Evaluation by Faculty Members of the Institute and assigning grade:

	Item	Marks (out of 100)
	Background of the project	05
	Technical work on  1. Experiment performed.  2. Mathematical modelling if any	
<b>D</b>	3. Design	30
Report	<ul><li>4. Techno-economic feasibility</li><li>5. Analysis of data</li></ul>	
	Conclusion	10
	Writing skills including formatting as per the given instructions	05
Presentation	<ol> <li>Presentation based on the work performed and its analysis.</li> <li>Presentation skill</li> </ol>	20
Industry mentor	Marks given by the industry mentor	30
To		

<sup>(</sup>a) The candidates who obtain 40% and more marks of the total marks of a subject head shall be deemed to have passed the respective subject head.

### Rules for assigning course codes:

### Core courses

<sup>(</sup>b) The candidates who obtain marks less than 40% of the total marks of a subject head shall be deemed to have failed in the respective subject head (Grade FF).

- o Course codes for Mathematics theory courses will start with MAT 22XX. Course codes for Statistics courses will start with MAT 232X. Course codes for Lab courses will start with MAP 252X.
- o In the revised syllabus some core courses are retained from the old syllabus with less than 25% changes in the syllabus. For these courses course codes remain unchanged. The codes are MAT 2210, MAT 2202, MAT 2207, MAT 2210, MAT 2206, MAP 2701.
- If a new code is given to an existing core course without any changes, equivalent codes are Approve by Academic Council on Aug. provided in a separate table. For continuity and maintaining uniformity, Project (SEM-IV) has