

**Ph.D Semester 1/2**  
**(CO / IT Engineering)**

**Subject Name:** Data Science for Engineers  
**Type of course:** PhD Course Work  
**Prerequisites (if any):** - Statistics

**Subject Code:** PTCO13104

**Rationale:** This course introduces the mathematical foundations required for data science. As part of this course, students will learn the framework for solving problems requiring data analytics. This course also introduces 'R' programming language. Students will be able to solve predictive problems on real time data after learning this course.

**Teaching and Examination Scheme:**

| Teaching Scheme |   |   |   | Theory Marks |     |     | Practical Marks |     | Total |
|-----------------|---|---|---|--------------|-----|-----|-----------------|-----|-------|
| L               | T | P | C | TEE          | CA1 | CA2 | TEP             | CA3 |       |
| 4               | 0 | 0 | 4 | 60           | 25  | 15  | 0               | 0   | 100   |

CA1: Continuous Assessment (assignments/projects/open book tests/closed book tests) CA2: Sincerity in attending classes/class tests/ timely submissions of assignments/self-learning attitude/solving advanced problems TEE: Term End Examination TEP: Term End Practical Exam (Performance and viva on practical skills learned in course) CA3: Regular submission of Lab work/Quality of work submitted/Active participation in lab sessions/viva on practical skills learned in course

**Contents**

| Sr No | Description   | No. of Hours |
|-------|---|--------------|
| 1.    | Course philosophy and expectation, Introduction to R, Variables and data types in R, Data frames, Recasting and joining of data frames, Arithmetic, Logical and Matrix operations in R, Advanced programming in R : Functions, Control structures, Data visualization in R Basic graphics | 8            |
| 2.    | Linear Algebra for Data science, Solving Linear Equations, Linear Algebra - Distance, Hyperplanes and Halfspaces, Eigenvalues, Eigenvectors   | 6            |
| 3.    | Statistical Modelling, Random Variables and Probability Mass/Density Functions, Sample Statistics, Hypotheses Testing   | 6            |
| 4.    | Optimization for Data Science, Unconstrained Multivariate Optimization, Gradient Descent, Learning Rule   | 6            |



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|    |  |    |
|----|--|----|
| 5. | Multivariate Optimization With Equality Constraints, Multivariate Optimization With Inequality Constraints, Introduction to Data Science, Solving Data Analysis Problems - A Guided Thought Process                        | 6  |
| 6. | Module: Predictive Modelling, Linear Regression, Model Assessment, Diagnostics to Improve Linear Model Fit, Simple Linear Regression Model Building, Simple Linear Regression Model Assessment, Multiple Linear Regression | 6  |
| 7. | Cross Validation, Multiple Linear Regression Modelling Building and Selection, Classification, Logistic Regression, Performance Measures, Logistic Regression Implementation in R  | 6  |
| 8. | K - Nearest Neighbors (kNN), K - Nearest Neighbors implementation in R, K - means Clustering, K - means implementation in R  | 6  |
| 9. | Applications of Data Science, Challenges reported in contemporary literature, Case Studies   | 10 |

**Course Outcomes**

| Sr. No. | CO Statement   | Marks % Weightage |
|---------|--|-------------------|
| 1.      | Describe a flow process for data science problems                            | 20%               |
| 2.      | Classify data science problems into standard typology                        | 20%               |
| 3.      | Develop R codes for data science solutions                                   | 20%               |
| 4.      | Correlate results to the solution approach                                   | 15%               |
| 5.      | Assess the solution approach   | 15%               |
| 6.      | Construct use cases to validate approach and identify modifications required | 10%               |

**Reference Books:**

| Sr no | Title of book /article                           | Author(s)          | Publisher            | Year of publication | Publication Edition |
|-------|--|--------------------|----------------------|---------------------|---------------------|
| 1     | Introduction To Linear Algebra                   | Gilbert Strang     | Wellesley Publishers | 2016                | Fifth Edition       |
| 2     | Applied Statistics And Probability For Engineers | Douglas Montgomery | Wiley                | 2016                | Sixth Edition       |





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|   |   |                              |  |      |             |
|---|---|------------------------------|--|------|-------------|
| 3 | Practical Statistics for Data Scientists                | Peter Bruce and Andrew Bruce | Publisher(s): O'Reilly Media, Inc. ISBN: 9781491952962 | 2017 | 2nd Edition |
| 4 | Think Stats: Probability and Statistics for Programmers | Allen B. Downey              | Green Tea Press<br>Needham,<br>Massachusetts           | 2014 | 2nd Edition |
| 5 | Data Science for Dummies                                | John Muller                  | Wiley  | 2017 | 2nd Edition |
| 6 | Statistics for Data Science                             | James D. Miller              | PACKT  | 2017 | -           |

**List of Online Learning Resources:**

3. NPTEL course on: Data Science for Engineers  
([https://onlinecourses.nptel.ac.in/noc23\\_cs17/preview](https://onlinecourses.nptel.ac.in/noc23_cs17/preview))

