

ವಿಶ್ವೇಶ್ವರಯ್ಯ ತಾಂತ್ರಿಕ ವಿಶ್ವವಿದ್ಯಾಲಯ



'ವಿ ಟಿ ಯು ಆಧಿನಿಯಮ ೧೯೯೪' ರ ಆಡಿಯಲ್ಲಿ ಕರ್ನಾಟಕ ಸರ್ಕಾರದಿಂದ ಸ್ವಾಪಿತವಾದ ರಾಜ್ಯ ವಿಶ್ವವಿದ್ಯಾಲಯ

Visvesvaraya Technological University

(State University of Government of Kamataka Established as per the VTU Act, 1994) "Inana Sangama" Belagavi-590018, Kamataka, India

Prof. B.E. Rangaswamy, Ph.D. Registrar

Ref. No.: VTU/BOS/A12/2024-25/5 66 9

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Circular

Sub: Updated Syllabus of MCS102 (Data Science & Management)

Ref: 1) Circular No. VTU/BGM/BoS/MCS101/691/2024-25

dated 5602 dated 29-01-2025

Feedback from concerned different engineering colleges

 Recommendations from University BOS Chairperon, CSE/ISE/MCA dated

With reference to the above, this is to inform you that based on the feedback received from concerned different engineering colleges and as per the recommendations from the University BOS Chairperon, CSE/ISE/MCA, it is informed to follow the existing Syllabus available on the website for the subject MCS102 — Data Science & Management (1st Semester common course for all PG Programmes under CSE Board).

You are hereby requested to bring the contents of this Circular to the notice of all the concerned faculty members/students of your college and inform them to follow the same.

 The Principals of all Engineering Colleges (Constituent, Non Autonomous and Autonomous) under VTU.

 The Chairpersons/Programme Coordinators of all VTU PG Centres at Muddenhalli, Belagavi, Mysuru and Kalaburagi Regions.

Copy to:

To

1) The Registrar (Eval.), VTU, Belagavi.

- 2) The Regional Director (I/c), VTU Regional Offices at Bengaluru, Belagavi, Kalaburagi & Mysuru for information and circulation.
- The Secretary to VC, VTU, Belagavi.
- 4) PS to Registrar, VTU, Belagavi.
- 5) Director, NAAC, VTU, Belagavi.



| | | Data Science and Ma | | Tara |
|--|--|--|--|---|
| Course Code | | MCS102 | CIE Marks | 50 |
| | urs/Week (L:P:SDA) | 4:0:2 | SEE Marks | 50 |
| Total Hours of | of Pedagogy | 50 | Total Marks | 100 |
| Credits | | 03 | Exam Hours | 03 |
| | | oncepts of data science, | including its history, signifi- | icance, and the d |
| 2. App | oly statistical methods a | and data analysis techniq | ues to interpret and draw in | sights from comp |
| datasets. | | | | |
| | | | and assess their performance | e using appropri |
| evaluation m | etrics in real-world scen | arios. | | |
| 4. Util | ize data visualization to | ools and techniques to e | ffectively communicate find | lings and insights |
| | | | | |
| diverse audie | nces. | | | |
| diverse audie | ences. | | | |
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Teaching Chalk and talk/PPT/case study/web content
Learning
Process
Module-5

Understanding Optimization Techniques, Typology of Data Science Problems, Solution Framework for Data

Module 4: Optimization and Data Science Problem Solving, Introduction to Optimization

Science Problems.

| Module 5: Regression | on and Classification Techniques, Linear Regression, Simple Linear Regression and |
|----------------------|---|
| Assumptions, Mul | ivariate Linear Regression, Model Assessment and Variable Importance, Subset |
| Selection, Classific | eation Techniques, Classification using Logistic Regression. |
| Teaching- | Chalk and talk/PPT/case study/web content |
| Learning | |
| Process | |

Assessment Details (both CIE and SEE)

The weightage of Continuous Internal Evaluation (CIE) is 50% and for Semester End Exam (SEE) is 50%. The minimum passing mark for the CIE is 50% of the maximum marks. Minimum passing marks in SEE is 40% of the maximum marks of SEE. A student shall be deemed to have satisfied the academic requirements and earned the credits allotted to each subject/course if the student secures not less than 50% (50 marks out of 100) in the sum total of the CIE (Continuous Internal Evaluation) and SEE (Semester End Examination) taken together.

Continuous Internal Evaluation:

- Three Unit Tests each of 20 Marks
- Two assignments each of 20 Marks or one Skill Development Activity of 40 marks to attain the COs and POs

The sum of three tests, two assignments/skill Development Activities, will be scaled down to 50 marks

CIE methods /question paper is designed to attain the different levels of Bloom's taxonomy as per the outcome defined for the course.

Semester End Examination:

- The SEE question paper will be set for 100 marks and the marks scored will be proportionately reduced to 50.
- The question paper will have ten full questions carrying equal marks.
- 3. Each full question is for 20 marks. There will be two full questions (with a maximum of four sub-questions) from each module.
- Each full question will have a sub-question covering all the topics under a module.
- 5. The students will have to answer five full questions, selecting one full question from each module

Suggested Learning Resources:

Text Books:

- Python for Data Analysis" by Wes McKinney, 2nd Edition (2018)
- "Data Science from Scratch: First Principles with Python" by Joel Grus, 2nd Edition (2019)

Reference Books:

- "An Introduction to Statistical Learning" by Gareth James, Daniela Witten, Trevor Hastie, and Robert Toshigami, 2nd Edition (2021)
- 2. "The Elements of Statistical Learning" by Trevor Hastie, Robert Toshigami, and Jerome Friedman, 2nd Edition (2009)
- "Data Science for Business: What You Need to Know about Data Mining and Data-Analytic Thinking" by Foster Provost and Tom Fawcett, 2nd Edition (2013)

Web links and Video Lectures (e-Resources):

https://www.coursera.org/specializations/jhu-data-science

https://www.kaggle.com/learn/data-science

https://www.edx.org/professional-certificate/harvardx-data-science

https://www.youtube.com/playlist?list=PL4cUxeGkcC9g1s4L6G8p8Fq5XK6Pq7b1k

| Sl. No. | Description | |
|---------|--|-----|
| 1 | Demonstrate the ability to independently conduct research and development work to address practical engineering problems. | PO1 |
| 2 | Develop and deliver comprehensive technical presentations that effectively convey complex information to diverse audiences. | PO2 |
| 3 | Exhibit mastery in the specialized study area, surpassing therequirements of a relevant bachelor's program. | PO3 |
| 4 | Analyze engineering problems critically and apply appropriatetechniques, skills, and modern tools to develop innovative solutions. | PO4 |
| 5 | Collaborate effectively in teams while also functioning independently, recognizing opportunities for career advancement andresearch. | PO5 |
| 6 | Cultivate a proactive approach to continuous learning and professional development in response to evolving technological landscapes. | PO6 |

Skill Development Activities Suggested

• The students with the help of the course teacher can take up relevant technical activities which will enhance their skill. The prepared report shall be evaluated for CIE marks.

Course outcome (Course Skill Set)

At the end of the course the student will be able to:

| Sl. No. | Description | Blooms |
|---------|-------------|--------|
| | | Level |

| | | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 |
|------------------------|-----|-----|-----|-----|-----|-----|-----|
| Mapping of COS and Pos | CO1 | х | | | x | + | |
| n ³ | CO2 | | | х | 1 | x | |
| | CO3 | | x | | | | |
| | CO4 | x | _ | | | | |

| CO1 | Explore the foundational concepts of data science, history, significance, and process. | | | | | |
|-----|--|-----|--|--|--|--|
| CO2 | Apply statistical methods and data analysis techniques to interpret and draw insights from complex datasets. | nL3 | | | | |
| CO3 | Implement various machine learning algorithms and assess their performance using appropriate evaluation metrics in real-world scenarios. | gL2 | | | | |
| CO4 | Utilize data visualization tools and techniques to effectively communicate findings and insights to diverse audiences. | iL4 | | | | |