

Programme and Course Structure

Sharda School of Basic Sciences & Research Department of Mathematics

B.Sc. (Hons./Hons. With Research) Data Science & Analytics

Programme Code: SBR0308

Batch 2024-28



Vision. Mission and Core Values of the University

Vision of the University

To serve the society by being a global University of higher learning in pursuit of academic excellence, innovation and nurturing entrepreneurship.

Mission of the University

M1. Transformative educational experience.

M2. Enrichment by educational initiatives that encourage global outlook.

M3. Develop research, support disruptive innovations and accelerateentrepreneurship.

M4. Seeking beyond boundaries.

Core Values

- 1. Integrity
- 2. Leadership
- 3. Diversity
- 4. Community



Vision and Mission of School

Vision of the School

Achieving excellence in the realm of science to address the challenges of evolving society.

Mission of the School

- 1. Equip the students with knowledge and skills.
- 2. Capacity building by providing academic flexibility to student and Faculty members.
- 3. To establish centre of excellence for innovative research.
- 4. Address the deficiencies of the society pertaining to environment
- 5. To strengthen academic- industry collaboration for better. Employability.
- 6. Developing a culture for continued betterment in all facets of life.

Core Values

- 1. Integrity
- 2. Leadership
- 3. Diversity
- 4. Community



Vision and Mission of Department of Mathematics

Vision of the Department

To become a globally recognized destination for education in applied mathematics and research.

Mission of the Department

1. To develop mathematical skills in students and make them employable across a wide range of professions and promote interest in research.

Core Values

- 2. To develop entrepreneurial skills in students to serve the society at large.
- 3. To develop skills for the applications of mathematics in the various fields.

- 1. Integrity
- 2. Leadership
- 3. Diversity
- 4. Community



B. Sc. (Hons./Hons. With Research) Data Science & Analytics

Programme Educational Objectives (PEOs)

PEO1: Prepare professionals conversant with current and advanced technological tools to carry out Investigation, analysis and synthesis by identifying various compute oriented solutions.

PEO2: To develop positive attitude and skills which enable them to become a multi facet personality.

PEO3: To prepare students in such a way so that they perform excellently in national label entrance examinations conducted by various well known institution like IIT's/ central Universities/other academic institutes etc. to pursue their PG/MS/Dual PG and Ph.D. programs.

PEO4: To make them aware of effective machine learning and Artificial Intelligence based data analytics and inference required for Industrial Application.

PEO5: To inculcate passion for lifelong learning by introducing principles of group dynamics, public policies, environmental and societal context.

Programme Outcomes (POs)

PO1. Complex Problem Solving: Solve different kinds of problems in familiar and non-familiar contexts and apply the learning to real-life situations.

PO2. Critical Thinking: Analyze and synthesize data from a variety of sources and draw valid conclusions and support them with evidence and examples.

PO3. Creativity: Demonstrate the ability to think 'out of the box' and generate solutions to complex problems in unfamiliar contexts by applying concepts of multidisciplinary and interdisciplinary.

PO4. Analytical reasoning/thinking: Evaluate the reliability and relevance of evidence.

PO5. Research-related skills: Demonstrate the ability to acquire the understanding of basic research ethics and skills in practicing/doing ethics in the field/ in personal research work, regardless of the funding authority or field of study.

PO6. Communication Skills: Demonstrate the skills that enable them to express thoughts and ideas effectively in writing and orally and communicate with others using appropriate media.

PO7. Coordinating/collaborating with others: Demonstrate the ability to work effectively and respectfully with diverse teams using management skills to guide people to the right destination.

PO8. Digital and technological skills: Demonstrate the capability to access, evaluate, and use a variety of relevant information sources, and use appropriate software for analysis of data.

PO9. Value Inculcation: Instill integrity and identify ethical issues related to work, and follow ethical practices with or understand the perspective, experiences, or points of view of another individual or group, and to identify and understand other people's emotions.

PO10. Sustainability Growth: Demonstrate the capability to lead a diverse team or individual to accomplish and participate in community-engaged services/ activities for promoting the well-being of society to mitigating the effects of environmental degradation, climate change, and pollution.

PO11. Multidisciplinary Life-long learning: Comprehensive knowledge and coherent understanding of the chosen disciplinary/interdisciplinary areas of study in a broad multidisciplinary context by inculcating a healthy attitude to be a lifelong learner,



Programme Specific Outcomes of B.Sc. (Hons./Hons. With Research) Data Science & Analytics

PSO1. Demonstrate proficiency with statistical analysis of data using advanced application tools.

PSO2. Apply data science concepts and methods to solve problems in real-world contexts.

PSO3. Estimate predictions for a given complex problem using data analytical methods not limited to machine learning and deep learning concepts.

| PEO Statements | School Mission1 | School Mission2 | School Mission3 | School Mission4 | School Mission5 | School Mission6 |
|-------------------|--------------------|--------------------|--------------------|--------------------|--------------------|--------------------|
| PEO1 | 3 | 2 | 3 | 1 | 2 | 3 |
| PEO2 | 3 | 2 | 3 | 1 | 2 | 3 |
| РЕОЗ | 3 | 3 | 3 | 3 | 3 | 3 |
| PEO4 | 3 | 2 | 3 | 1 | 3 | 3 |
| PEO5 | 3 | 2 | 2 | 3 | 2 | 3 |

Mapping of PEOs with Mission Statements

Mapping of Programme Outcomes Vs Programme Educational Objectives

| | PEO1 | PEO2 | PEO3 | PEO4 | PEO5 |
|------|------|------|------|------|------|
| PO1 | 3 | 3 | 3 | 2 | 1 |
| PO2 | 3 | 3 | 3 | 2 | 1 |
| PO3 | 3 | 3 | 3 | 2 | 1 |
| PO4 | 3 | 2 | 3 | 2 | 1 |
| PO5 | 2 | 3 | 2 | 3 | 2 |
| PO6 | 3 | 3 | 3 | 2 | 2 |
| PO7 | 1 | 2 | 1 | 3 | 1 |
| PO8 | 2 | 2 | 1 | 3 | 2 |
| PO9 | 2 | 2 | 2 | 3 | 2 |
| PO10 | 2 | 2 | 2 | 3 | 3 |
| PO11 | 2 | 3 | 3 | 2 | 1 |
| PSO1 | 2 | 1 | 1 | 2 | 1 |
| PSO2 | 2 | 2 | 2 | 2 | 1 |
| PSO3 | 3 | 3 | 3 | 2 | 1 |

1. Slight (Low) 2. Moderate (Medium) 3. Substantial (High)



4-Year Course Structure of B. Sc. (Hons./Hons. With Research) Data Science & Analytics

| | | Subject 1 | Subject 2 | Subj | ect 3 | Subject 4 | Vocational | Co- curricular | Training/Survey/ Project/ | | |
|------|-----|---------------------|--|------------------|--------------------|------------------------------|---|-----------------------------|---|---------------------------------|--|
| | | Major 1 | Major 2 | Maj | jor 3 | Minor/ Elective | Minor | Minor | Major | | {Cumulative |
| | | Credits (3/4//5) | Credits (3/4/5) | Credits (3/4//5) | Project | Credits (3/4/5) | Credits (3) | Credits (2) | Credits (2/3/4/8) | {Minimum Credits} For the | Minimum Credits} Required for Award of |
| | | CC | CC | DSI | Ξ | OPE | SEC | AEC | VAC | year | Certificate/ Diploma/ Degree |
| Year | | Own Faculty | Own/inter or multidisciplinary Faculty | Any Faculty | | Other subject/ Faculty | Vocational/Skill Development Course | Co- curricular course | Inter/Intra Faculty related to main Subject | | Dipionial Degree |
| | Ι | Maths (4+1) | Х | Stats (3) | Х | CS (3+1) | SEC (3) | SK (2) | VAC (3) | | {40} |
| 1 | II | Maths (4+1) | CS (3+1) | Х | Х | Stats (3) | SEC (3) | SK (2) | VAC (3) | {40} | Certificate in Faculty |
| 2 | III | Stats (4+1) | CS (3+1) | Maths (4) | RBL-1** (Audit) | Maths (3) | SEC (3) | SK (2) | Х | {40} | {80} |
| 2 | IV | CS (3+1) | CS (4+1) | Stats (4+1) | RBL-2** (Audit) | OPE-1(3) | Х | SK (2) | Х | <u>ر</u> مير | Diploma in Faculty |

Department of Mathematics

** Courses are the audit courses. However, evaluation shall be made as per rubrics. Summer Industry Internship (Industry Connect)

Course shall be conducted in the summer break of 04th Semester. However, evaluation will be made as per Rubrics in the 5th Semester "Industry Connect"

[^]Maths-Mathematics course; Stats-Statistics course; CS-Computer Science course; SK-Sharda Skills course; SEC-Skill Enhancement Course; AEC-Ability Enhancement Course; VAC-Value Added Course; OPE-Open Elective course.



| 3 | V | CS (5) CS (4+1) | Stats (3+1) | Stats (2+1) | RBL-3 (1) | х | Х | Х | Industry Connect (2) | {40} | {120} |
|---|------|--|-------------|--------------------------|--------------|-----------|---|---|--------------------------|-------------------|---|
| 3 | VI | Maths (4+1) Stats (4+1) | Stats (3+1) | Х | RBL-4 (1) | CS (3) | Х | Х | Community Connect (2) | | Bachelor Degree inFaculty |
| 4 | VII | CS (4) CS (4) | Х | Stats (3+1) Stats (4) | Х | OPE-2 (4) | Х | Х | Х | {40} | {160} |
| 4 | VIII | CS (4) CS (4) Stats (4) | Х | Stats (4) | Х | OPE-3 (4) | Х | Х | Х | { 4 0} | Bachelor (Hons.) in Faculty |
| 4 | VII | CS (4) CS (4) Stats (3+1) Stats (4) | Х | Х | Х | OPE-2 (4) | Х | Х | Dissertation-1 (3) | {40} | {160} |
| 4 | VIII | CS/Stats (4) | Х | Х | Х | OPE-3 (4) | Х | Х | Dissertation-2 (9) | With Resear | Bachelor (Hons. With Research) in Faculty |

[^]Maths-Mathematics course; Stats-Statistics course; CS-Computer Science course; SK-Sharda Skills course; SEC-Skill Enhancement Course; AEC-Ability Enhancement Course; VAC-Value Added Course; OPE-Open Elective course.



Programme Structure Template B. Sc. (Hons./Hons. With Research) Data Science & Analytics Term: 2401 (Semester-I)

Batch: 2024-28

| S. No. | Course Code | Course Name | Teaching Load | | | ıd | Credits | Pre-Requisite/ Co-Requisite | Type of Course: 1. CC; 2. DSE; 3. OPE; 4. SEC; 5. AEC; 6. VAC; 7. Project |
|--------|-------------|---|---------------|---|---|----------------|---------|---------------------------------|---|
| | THEORY | | L | Т | Р | TOTAL (hrs) | | | |
| 1. | MSM101 | Foundation Course in Mathematics | 4 | 0 | 0 | 4 | 4 | Basic Mathematics up to 10+2 | CC |
| 2. | CMS102 | Descriptive Statistics | 3 | 0 | 0 | 3 | 3 | Basic Mathematics up to 10+2 | OPE |
| 3. | CSE113 | Programming for Problem Solving | 3 | 0 | 0 | 3 | 3 | | DSE (Multi/Inter-discpli) |
| 4. | VAC103 | Environmental Management | 3 | 0 | 0 | 3 | 3 | | VAC |
| | PRACTICALS | | | | | | | | |
| 5. | ARP101 | Communicative English-1 | 1 | 0 | 2 | 3 | 2 | | AEC |
| 6. | | Essential Excel Skills for Business | 0 | 0 | 6 | 6 | 3 | | SEC |
| 7. | CMS151 | Foundation Course in Mathematics Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite MSM101 | CC |
| 8. | CSP113 | Programming for Problem Solving Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite CSE113 | DSE (Multi/Inter-discpli) |
| | | TOTAL CREDITS | | | | | 20 | | |



| | | B. Sc. (Hons./Hons. With | | <i>,</i> | | • | ics | Batch: 2024-28 | |
|--------|-------------|--|---|----------|---|----------------|---------|--------------------------------|--|
| S. No. | Course Code | Course Name | TERM: 2402 (Semester-II) Teaching Load | | | | Credits | Pre-Requisite/ Co-Requisite | Type of Course: 1. CC; 2. DSE; 3. OPE; 4. SEC; 5. AEC; 6. VAC; 7.Project |
| | THEORY | | L | Т | Р | TOTAL (hrs) | | | |
| 1. | CMS131 | Matrix Analysis and Linear Algebra | 4 | 0 | 0 | 4 | 4 | Pre-requisite MSM101 | CC |
| 2. | CMS132 | Mathematical Expectations & Probability Distributions | 3 | 0 | 0 | 3 | 3 | Pre-requisite CMS102 | OPE |
| 3. | CSE242 | Data Structures | 3 | 0 | 0 | 3 | 3 | Pre-requisite CSE113 | CC |
| | PRACTICALS | | | | | | | | |
| 4. | ARP102 | Communicative English-2 | 1 | 0 | 2 | 3 | 2 | Pre-requisite ARP101 | AEC |
| 5. | VOM104 | Advanced Excel Skills for Business | 0 | 0 | 6 | 6 | 3 | Pre-requisite VOM103 | SEC |
| 6. | VAC110 | Yoga for Holistic Health | 0 | 1 | 4 | 5 | 3 | | VAC |
| 7. | CMS171 | Matrix Analysis and Linear Algebra Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite CMS131 | CC |
| 8. | CSP242 | Data Structures Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite CSE113 | CC |
| | | TOTAL CREDITS | | | | | 20 | | |

Programme Structure Template



| | B. | Sc. (Hons./Hons. With Researc | , | Science | & Ana | lytics | | Batch: 2024-28 | |
|--------|-------------|---|--|---------|-------|----------------|---------|--------------------------------|--|
| S. No. | Course Code | Course Name | TERM: 2501 (Semester-III) Teaching Load | | | | Credits | Pre-Requisite/ Co-Requisite | Type of Course: 1. CC; 2. DSE; 3. OPE; 4. SEC; 5. AEC; 6. VAC; 7.Project |
| | THEORY | | L | Т | Р | TOTAL (hrs) | | | |
| 1. | MSM312 | Discrete Mathematics | 3 | 1 | 0 | 4 | 4 | Pre-requisite MSM101 | DSE |
| 2. | BDA215 | Operations Research | 3 | 0 | 0 | 3 | 3 | Pre-requisite MSM101 | OPE |
| 3. | BDA216 | Statistical Inference | 4 | 0 | 0 | 4 | 4 | Pre-requisite CMS132 | CC |
| 4. | BDA217 | Data Preparation and Data Cleaning | 3 | 0 | 0 | 3 | 3 | Pre-requisite CMS132 | CC |
| | PRACTICALS | | | | | | | | |
| 5. | VOM203 | Basic Excel Modelling | 0 | 0 | 6 | 6 | 3 | Pre-requisite VOM104 | SEC |
| 6. | ARP207 | Logical Skill Building & Soft Skills | 0 | 1 | 2 | 3 | 2 | Pre-requisite ARP102 | AEC |
| 7. | BDA261 | Statistical Inference Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite BDA216 | CC |
| 8. | BDA262 | Data Preparation and Data Cleaning Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite BDA217 | CC |
| 9. | RBL001 | Research Report Writing-I (RBL-1) | 0 | 0 | 2 | 2 | 0 | Pre-requisite ARP102 | Project (Non-graded Qualifying) |
| | | TOTAL CREDITS | | | | | 21 | | |

Programme Structure Template



Programme Structure Template B. Sc. (Hons./Hons. With Research) Data Science & Analytics TERM: 2502 (Semester-IV)

Batch: 2024-28

| S. No. | Course Code | Course Name | Teaching Load | | | | Credits | Pre-Requisite/ Co-Requisite | Type of Course: 1. CC; 2. DSE; 3. OPE; 4. SEC; 5. AEC; 6. VAC; 7.Project |
|--------|-------------|--|---------------|---|---|----------------|---------|--------------------------------|--|
| | THEORY | | L | Т | Р | TOTAL (hrs) | | | |
| 1. | BDA218 | Data Ware Housing & Data Mining | 3 | 0 | 0 | 3 | 3 | Pre-requisite BDA217 | CC |
| 2. | BDA202 | Database Management Systems | 4 | 0 | 0 | 4 | 4 | Pre-requisite MSM312 | CC |
| 3. | BDA214 | Sampling Theory | 4 | 0 | 0 | 4 | 4 | Pre-requisite BDA216 | DSE |
| 4. | OPE | Open Elective-1 | 3 | 0 | 0 | 3 | 3 | | OPE |
| | PRACTICALS | | | | | | | | |
| 5. | ARP306 | Campus to Corporate | 0 | 1 | 2 | 3 | 2 | Pre-requisite ARP207 | AEC |
| 6. | BDA270 | Data Ware Housing & Data Mining Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite BDA218 | CC |
| 7. | BDA271 | Database Management Systems Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite BDA202 | CC |
| 8. | BDA272 | Sampling Theory Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite BDA214 | DSE |
| 9. | RBL002 | Research Based Learning-II (RBL-2) | 0 | 0 | 2 | 2 | 0 | Pre-requisite RBL001 | Project (Non-graded Qualifying) |
| | | TOTAL CREDITS | | | | | 19 | | |



Programme Structure Template B. Sc. (Hons./Hons. With Research) Data Science & Analytics TERM: 2601 (Semester-V)

Batch: 2024-28

| S. No. | Course Code | Course Name | Teaching Load | | | | Credits | Pre-Requisite/ Co-Requisite | Type of Course: 1. CC; 2. DSE; 3. OPE; 4. SEC; 5. AEC; 6. VAC; 7.Project |
|--------|-------------------|--|---------------|---|---|----------------|---------|--------------------------------|--|
| | THEORY | | L | Т | Р | TOTAL (hrs) | | | |
| 1. | BDA346 | Artificial Intelligence | 5 | 0 | 0 | 5 | 5 | Pre-requisite BDA218 | CC |
| 2. | BDA303 | Machine Learning | 4 | 0 | 0 | 4 | 4 | Pre-requisite BDA218 | CC |
| 3. | BDA319 | Regression Analysis | 3 | 0 | 0 | 3 | 3 | Pre-requisite BDA214 | CC |
| 4. | BDA320/ BDA321 | Advanced Statistical Analysis/ Experimental Design | 2 | 0 | 0 | 2 | 2 | | DSE (Multi/Inter-discpli) |
| | PRACTICALS | | | | | | | | |
| 5. | BDA355 | Machine learning Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite BDA303 | CC |
| 6. | BDA356 | Regression Analysis Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite BDA319 | CC |
| 7. | INC001 | Industry Connect | 0 | 0 | 4 | 4 | 2 | | Project |
| 8. | RBL003 | Research Based Learning-III (RBL-3) | 0 | 0 | 2 | 2 | 1 | Pre-requisite RBL002 | Project |
| 9. | BDA359/ BDA363 | Advanced Statistical Analysis Lab/ Experimental Design Lab | 0 | 0 | 2 | 2 | 1 | | DSE (Multi/Inter-discpli) |
| | | TOTAL CREDITS | | | | | 20 | | |



Programme Structure Template B. Sc. (Hons./Hons. With Research) Data Science & Analytics TERM: 2602 (Semester-VI)

Batch: 2024-28

| S. No. | Course Code | Course Name | Teaching Load | | | | Credits | Pre-Requisite/ Co-Requisite | Type of Course: 1. CC; 2. DSE; 3. OPE; 4. SEC; 5. AEC; 6. VAC; 7.Project |
|--------|-------------|---------------------------------------|---------------|---|---|----------------|---------|--------------------------------|--|
| | THEORY | | L | Т | Р | TOTAL (hrs) | | | |
| 1. | CMS331 | Numerical Methods | 4 | 0 | 0 | 4 | 4 | Pre-requisite CMS131 | CC |
| 2. | BDA322 | Statistical Simulation | 4 | 0 | 0 | 4 | 4 | Pre-requisite BDA319 | CC |
| 3. | BDA323 | Multivariate Data Analysis | 3 | 0 | 0 | 3 | 3 | Pre-requisite BDA319 | CC |
| 4. | BDA325 | Deep Learning | 3 | 0 | 0 | 3 | 3 | Pre-requisite BDA303 | OPE |
| | PRACTICALS | | | | | | | | |
| 5. | CMS371 | Numerical Methods Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite CMS331 | CC |
| 6. | BDA360 | Statistical Simulation Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite BDA322 | CC |
| 7. | BDA361 | Multivariate Data Analysis Lab | 0 | 0 | 2 | 2 | 1 | Co-requisite BDA323 | CC |
| 8. | CCU108 | Community Connect | 0 | 0 | 4 | 4 | 2 | | Project (Multi/Inter-discpli) |
| 9. | RBL004 | Research Based Learning-IV (RBL-4) | 0 | 0 | 2 | 2 | 1 | Pre-requisite RBL003 | Project |
| | | TOTAL CREDITS | | | | | 20 | | |

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Programme Structure Template B. Sc. (Hons./Hons. With Research) Data Science & Analytics TERM: 2701 (Semester-VII)

Batch: 2024-28

| S. No. | Course Code | Course Name | Teaching Load | | | | Credits | Pre-Requisite/ Co-Requisite | Type of Course: 1. CC; 2. DSE; 3. OPE; 4. SEC; 5. AEC; 6. VAC; 7.Project |
|--------|-------------------|---|---------------|---|---|----------------|---------|-----------------------------------|--|
| | THEORY | | L | Т | Р | TOTAL (hrs) | | | |
| 1. | MDA104 | Next Generation Databases | 4 | 0 | 0 | 4 | 4 | Pre-requisite BDA346, 303, 323 | CC |
| 2. | MDA109 | Big Data Analytics | 4 | 0 | 0 | 4 | 4 | Pre-requisite BDA323 | CC |
| 3. | | Time Series, Forecasting and Index Number/ Econometrics | 3 | 0 | 0 | 3 | 3 | | DSE/CC* |
| 4. | MDATTI/ MDATT3 | Non-Parametric Statistical Inference/ Survival Analysis | 4 | 0 | 0 | 4 | 4 | | DSE/CC* |
| 5. | OPE | Open Elective-1 | 4 | 0 | 0 | 4 | 4 | | OPE |
| | PRACTICALS | | | | | | | | |
| 6. | N/II) A I 56 | Time Series, Forecasting and Index Number Lab/ Econometrics Lab | 0 | 0 | 2 | 2 | 1 | | DSE/CC* |
| | | TOTAL CREDITS | | | | | 20 | | |

*Credited Research Project/Dissertation: Students of B.Sc. Data Science & Analytics have the option to choose a research project/dissertation of worth 12 credits (However student has to take 23 credits courses including 3 credits project in 7th semester and 17 credits courses including 9 credits project in 8th semester). This can be undertaken for those who secure 75% and above marks in the first six semesters and wish to undertake research at the undergraduate level can choose a research stream in the fourth year. They should do a research project or dissertation under the guidance of a mathematics faculty member of the Sharda University. The students, who secure 160 credits, including 12 credits from a research project/dissertation, are awarded **B.Sc. (Hons. with Research)** Data Science & Analytics.



Batch: 2024-28

| Programme Structure Template |
|---|
| B. Sc. (Hons./Hons. With Research) Data Science & Analytics |
| TERM: 2702 (Semester-VIII) |

Type of Course: 1. CC; 2. DSE; **Pre-Requisite**/ 3. OPE; 4. SEC; S. No. **Course Code Course Name Teaching Load** Credits **Co-Requisite** 5. AEC; 6. VAC; 7.Project TOTAL THEORY L Т Р (hrs) Advanced Big Data and Text Pre-requisite CC **MDA107** 4 0 0 1. 4 4 MDA109 Analytics Pre-requisite CC **MDA114** Bayesian Data Analysis 2. 4 0 0 4 4 BDA322, 323 Pre-requisite MDA117 Computational Intelligence CC 4 3. 0 0 4 4 BDA303.322.346. Demography/ MDA115/ DSE/CC* 4 0 0 4 4 4. MDA116 Statistical Quality Control Open Elective-2 OPE OPE 5. 4 0 0 4 4 **TOTAL CREDITS** 20

*Credited Research Project/Dissertation: Students of B.Sc. Data Science & Analytics have the option to choose a research project/dissertation of worth 12 credits (However student has to take 23 credits courses including 3 credits project in 7th semester and 17 credits courses including 9 credits project in 8th semester). This can be undertaken for those who secure 75% and above marks in the first six semesters and wish to undertake research at the undergraduate level can choose a research stream in the fourth year. They should do a research project or dissertation under the guidance of a mathematics faculty member of the Sharda University. The students, who secure 160 credits, including 12 credits from a research project/dissertation, are awarded B.Sc. (Hons. with Research) Data Science & Analytics.



| Sem | CC | DSE | OPE | SEC | AEC | VAC | Project | Mathematics | Computer Science | Statistics |
|--------|----|-------|-------|------|------|------|---------|-------------|-------------------------|------------|
| 1 | 4 | 4 | 4 | 3 | 2 | 3 | 0 | 4 | 4 | 4 |
| 2 | 8 | 0 | 4 | 3 | 2 | 3 | 0 | 4 | 4 | 4 |
| 3 | 8 | 3 | 4 | 3 | 2 | 0 | 0 | 3 | 8 | 4 |
| 4 | 9 | 5 | 4 | 0 | 2 | 0 | 0 | 0 | 9 | 9 |
| 5 | 14 | 3 | 0 | 0 | 0 | 0 | 3 | 0 | 10 | 7 |
| 6 | 17 | 0 | 0 | 0 | 0 | 0 | 3 | 5 | 3 | 9 |
| Total: | 60 | 15 | 16 | 9 | 8 | 6 | 6 | 16 | 38 | 37 |
| % | 50 | 12.5 | 13.33 | 7.5 | 6.67 | 5 | 5 | 13.33 | 31.67 | 30.83 |
| 7 | 8 | 8 | 4 | 0 | 0 | 0 | 0 | 0 | 8 | 8 |
| 8 | 12 | 4 | 4 | 0 | 0 | 0 | 0 | 0 | 3 | 13 |
| Total: | 80 | 27 | 24 | 9 | 8 | 6 | 6 | 16 | 49 | 58 |
| % | 50 | 16.88 | 15 | 5.63 | 5 | 3.75 | 3.75 | 10 | 30.63 | 36.25 |

B. Sc. (Hons./Hons. With Research) Data Science & Analytics Curriculum Credits Distribution



COURSE ARTICULATION MATRIX

| COs | PO 1 | PO 2 | PO 3 | PO 4 | РО 5 | PO 6 | PO 7 | PO 8 | PO 9 | PO 10 | PO 11 | PSO 1 | PSO 2 | PSO 3 |
|--------|---------|---------|---------|---------|---------|---------|----------------|---------|---------|----------|----------|----------|----------|----------|
| MSM101 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | | | |
| CMS102 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | 1.0 | | |
| CSE113 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | | | |
| VOM103 | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 1.0 | 1.0 | 1.0 | |
| ARP101 | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 1.0 | | | |
| VAC103 | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 2.0 | | | |
| CMS151 | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 1.0 | 1.0 | | |
| CSP113 | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 2.0 | 1.0 | | |
| CMS131 | 2.5 | 2.5 | 2.5 | 2.6 | | 1.0 | | | | | 1.0 | | 1.0 | |
| CMS132 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | 1.0 | 1.0 | |
| CSE242 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | | | |
| VOM104 | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | | | 1.0 | 1.0 | |
| ARP102 | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | | | | |
| VAC110 | 2.0 | | | 1.0 | 1.0 | | 2.0 | 2.0 | 1.0 | | | | | |
| CMS171 | 3.0 | 2.2 | 2.7 | 2.7 | 2.5 | 2.5 | 3.0 | 2.3 | 2.0 | 1.0 | 1.0 | 2.0 | 1.0 | |
| CSP242 | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 1.0 | | | |
| MSM312 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | | | | |
| BDA215 | 1.0 | 2.0 | 2.3 | 2.1 | 1.0 | 1.0 | 1.0 | 2.0 | 3.0 | | 1.0 | | | 1.0 |
| BDA216 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | 1.0 | 1.0 | 1.0 |
| RBL001 | 2.3 | 2.6 | 2.0 | 2.1 | 1.0 | 1.0 | | | | | 2.0 | 1.0 | | |
| BDA217 | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 2.0 | | 1.0 | 1.0 | |
| VOM203 | 2.0 | 1.0 | 2.0 | | 1.0 | 2.0 | 3.0 | 2.0 | | 2.0 | | | 1.0 | |
| ARP207 | 2.0 | 2.0 | 1.0 | 3.0 | 1.0 | 3.0 | 1.0 | | | | | | | |
| BDA261 | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 1.0 | 1.0 | 2.0 | |
| BDA262 | 1.0 | 2.0 | 2.0 | 2.0 | 1.0 | 1.0 | 1.0 | 3.0 | 1.0 | 1.0 | 3.0 | 1.0 | 2.0 | 3.0 |
| BDA218 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | 1.0 | | |
| BDA202 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | 1.0 | | |
| BDA214 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | 1.0 | 1.0 | |
| RBL002 | 2.3 | 2.6 | 2.0 | 2.1 | 1.0 | 1.0 | | | | | 1.0 | 1.0 | | |
| ARP306 | | | 2.0 | 2.0 | | 3.0 | 1.0 | 3.0 | 1.0 | | 2.0 | | | |
| BDA270 | 1.0 | 2.0 | 2.0 | 2.0 | 2.0 | 1.0 | 1.0 | 3.0 | 1.0 | 1.0 | 2.0 | 1.0 | 2.0 | 2.0 |
| BDA271 | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 2.0 | | 2.0 | |
| BDA272 | 1.0 | 2.0 | 2.0 | 2.0 | 1.0 | 1.0 | 1.0 | 3.0 | 1.0 | | 2.0 | 1.0 | | |
| BDA346 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 3.0 | 3.0 | | |
| BDA303 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 3.0 | 3.0 | | |
| BDA319 | 2.3 | 2.6 | 2.0 | 2.1 | 1.0 | 1.0 | | | | | 3.0 | 3.0 | | |
| BDA320 | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 3.0 | 3.0 | 1.0 | |
| BDA321 | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 1.0 | 1.0 | | |
| RBL003 | | 2.0 | 1.0 | 2.0 | 2.0 | 1.0 | | 3.0 | | | 2.0 | 2.0 | 2.0 | 2.0 |
| INC001 | 2.0 | 1.0 | 2.0 | 2.0 | 1.0 | 2.0 | 3.0 | 2.0 | 2.0 | 2.0 | 2.0 | 2.0 | 2.0 | |

| COs | PO | PO | PO | PC | | | 0 | P | | PO | PO | | PO | PO | PSO | PSO | PSO |
|--------|-----|-----|-----|-----|-----|-----|----|----|-----|-----|-----|----|-----|-----|-----|-----|-----|
| 0.03 | 1 | 2 | 3 | 4 | 5 | | 6 | 7 | 7 | 8 | 9 | | 10 | 11 | 1 | 2 | 3 |
| BDA355 | 1.0 | 2.0 | 2.0 | 2.0 | 2.0 |) 1 | 0. | 1. | .0 | 3.0 | 1.0 | | 2.0 | 3.0 | 2.0 | 2.0 | 3.0 |
| BDA356 | 1.0 | 2.0 | 2.0 | 2.0 | 2.0 | 1.0 | 1. | 0 | 3.0 | 1.0 | 2 | .0 | 2.0 | 2.0 | 2.0 | 3.0 | |
| BDA357 | 1.0 | 2.0 | 2.0 | 2.0 | 1.0 | 1.0 | 1. | 0 | 3.0 | 1.0 | 1 | .0 | 3.0 | 1.0 | 2.0 | 3.0 | |
| CMS331 | 2.5 | 2.5 | 2.5 | 2.6 | 2.0 | 1.0 | | | | | | | 2.0 | | | | |
| BDA322 | 2.3 | 2.6 | 2.0 | 2.1 | 2.0 | 1.0 | | | | | | | 2.0 | 2.0 | | | |
| BDA323 | 2.3 | 2.6 | 2.0 | 2.1 | 2.0 | 1.0 | | | | | | | 2.0 | 2.0 | | | |
| RBL004 | | | | 2.0 | 3.0 | 3.0 | 3. | 0 | 3.0 | 3.0 | 3 | .0 | 3.0 | 3.0 | 1.0 | 1.0 | |
| BDA324 | | 2.0 | 1.0 | 2.0 | | 1.0 | | | 3.0 | | | | 3.0 | 3.0 | | | |
| CCU108 | 1.0 | 2.0 | 2.0 | 3.0 | 3.0 | 2.0 | 3. | 0 | 3.0 | 3.0 | 2 | .0 | 1.0 | 2.0 | 2.0 | 3.0 | |
| CMS371 | 2.2 | 3.0 | 2.2 | 2.7 | 2.7 | 2.5 | 2. | 5 | 2.5 | 2.3 | 2 | 0 | 2.0 | | 1.0 | | |
| BDA360 | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1. | 0 | 3.0 | 1.0 |) | | 3.0 | 1.0 | 2.0 | 3.0 | |
| BDA361 | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1. | 0 | 3.0 | 1.0 |) | | 3.0 | 1.0 | 2.0 | 3.0 | |
| MDA104 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | | | 3.0 | | 3.0 | | |
| MDA109 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | | | 3.0 | | 3.0 | | |
| MDA110 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | | | 3.0 | | 3.0 | | |
| MDA111 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | | | 3.0 | | 3.0 | | |
| MDA112 | | 2.0 | 1.0 | 2.0 | | 1.0 | | | 3.0 | | | | 3.0 | 1.0 | 1.0 | 1.0 | |
| MDA113 | | 2.0 | 1.0 | 2.0 | | 1.0 | | | 3.0 | | | | 3.0 | | 1.0 | 1.0 | |
| MDA155 | | | | 2.0 | 1.0 | 1.0 | 1. | 0 | 3.0 | 1.0 | 1 | .0 | 3.0 | 1.0 | 2.0 | 3.0 | |
| MDA156 | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1. | 0 | 3.0 | 1.0 | | | 3.0 | 1.0 | 2.0 | 3.0 | |
| MDA107 | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | | | 3.0 | | | | |
| MDA114 | | 2.0 | 1.0 | 2.0 | | 1.0 | | | 3.0 | | | | 3.0 | | 1.0 | 1.0 | |
| MDA117 | | 2.0 | 1.0 | 2.0 | | 1.0 | | | 3.0 | | | | 3.0 | | 1.0 | 1.0 | |
| MDA115 | | 2.0 | 1.0 | 2.0 | | 1.0 | | | 3.0 | | | | 3.0 | 3.0 | 3.0 | | |
| MDA116 | | 2.0 | 1.0 | 2.0 | | 1.0 | | | 3.0 | | | | 3.0 | 3.0 | 3.0 | | |

1. Slight (Low) 2. Moderate (Medium) 3. Substantial (High)

A+



Detailed Syllabus for

CERTIFICATE COURSE IN

APPLIED MATHEMATICS



| Scho | ol: SSBSR | Batch: 2024-28 | | | | | | | |
|------|-------------------------------|--|--|--|--|--|--|--|--|
| | ramme: B.Sc. | Academic Year: 2024-25 | | | | | | | |
| (Hor | | | | | | | | | |
| | ich: Data Science nalytics | Semester: 1 | | | | | | | |
| 1 | Course Code | MSM101 | | | | | | | |
| 2 | Course Title | Foundation Course in Mathematics | | | | | | | |
| 3 | Credits | 4 | | | | | | | |
| 4 | Contact Hours (L-T-P) | 4-0-0 | | | | | | | |
| | Course Status | CC | | | | | | | |
| 5 | Course Objective | To familiarise the students with basic concepts of matrices, dete solving the system of linear equations. To understand the basic concept of sets theory, co-ordinate geom number, and vector algebra. | | | | | | | |
| 6 | Course Outcomes | CO1: Explain the concept of matrices and solve systems of linear equideterminants. (K2, K3, K4) CO2: Explain the concept of complex numbers and calculate the nth riscomplex numbers and illustrate the solutions of simple Polynomial equides (K3, K4) | oots of | | | | | | |
| | | CO3: Memorize the basic of Cartesian coordinate system and use alg techniques to explain intercepts and explore equations of lines on the plane. (K1, K3, K4) | ues to explain intercepts and explore equations of lines on the number | | | | | | |
| | | CO4: Describe and differentiate the symmetries from graphs of conic (K1, K2) | sections. | | | | | | |
| | | CO5: Describe and use the concepts of set theory, relation and function K3) | x · · · | | | | | | |
| | | CO6: Explain the basic concepts of vector algebra and use to parallelogram and quadrilateral, Vector triple product. (K2, K3, K4) | find area of | | | | | | |
| 7 | Course Description | This course is an introduction to the fundamental of Mathematics. objective of the course is to develop the basic understanding of lin complex number, co-ordinate geometry, sets theory and vector algeb | near algebra, | | | | | | |
| 8 | Outline syllabus | | CO Mapping | | | | | | |
| | Unit 1 | Matrices | •• • • | | | | | | |
| | А | Evaluation of determinants, Properties of determinants, | CO1 | | | | | | |
| | В | Matrices: types of matrices, addition, subtraction and multiplication of matrices, symmetric and skew-symmetric matrix. Inverse of matrix. | CO1 | | | | | | |
| | С | Rank of a matrix, Consistency of system of equations, Characteristic equation, Cayley -Hamilton theorem. | CO1 | | | | | | |
| | Unit 2 | Complex Numbers | | | | | | | |
| | А | Representation of complex number in Argand plane, Modulus and argument of complex number | CO2 | | | | | | |
| | В | Algebraic operations, De- Moivre's theorem | CO2 | | | | | | |
| | С | Nth root of complex number, Euler's formula | CO2 | | | | | | |
| | Unit 3 A | Co-ordinate geometry Cartesian coordinate system, Distance between two points Equations of line in various forms | CO3 | | | | | | |
| | В | Equations of line in various forms Equation of circle in various forms, Equation of tangent and normal to the circle. | | | | | | | |
| | С | Equation of ellipse, parabola and hyperbola | CO3, CO4 | | | | | | |
| | Unit 4 | Set Theory | , | | | | | | |



| А | Definition of set, types of sets, Union and the intersection of sets, Venn diagram, De-Morgan's law. | CO5 |
|---------------------------|---|-----|
| В | Relation and functions. | CO5 |
| С | Composite function and inverse function. | CO5 |
| Unit 5 | Vector Algebra | |
| А | Addition and subtraction of vectors and their geometric application. | CO6 |
| В | Scalar and vector product, their physical application, Projection of vector on another vector, area of the triangle. | CO6 |
| С | Area of parallelogram and quadrilateral, Vector triple product. | CO6 |
| Mode of | Theory | |
| examination | | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | Kreyszig, E., "Advanced Engineering Mathematics", John Wiley & Sons Inc. Jain, M.K., and Iyengar, S.R.K., "Advanced Engineering Mathematics", Narosa Publications | |
| Other References | Thomas, B.G., and Finny R.L., "Calculus and Analytical geometry", Pearson Education Asia, AdisonWisley. Simmons, G.F., "Differential Equations with applications with applications", Tata McGraw-Hill. | |

| PO | PO | PO | PO | PO | PO | PO | PO | PO | PO | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MSM101.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | | | |
| MSM101.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 1 | | | |
| MSM101.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 1 | | | |
| MSM101.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 1 | | | |
| MSM101.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | | | |
| MSM101.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 1 | | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | | | |



| | ool: SSBSR | Batch: 2024-28 | | | | | | | | |
|-----|---|--|---|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2024-25 | | | | | | | | |
| (Ho | | | | | | | | | | |
| | nch: Data Science | Semester: I | | | | | | | | |
| | nalytics | CMG100 | | | | | | | | |
| 1 | Course Code | CMS102 | | | | | | | | |
| 2 | Course Title | Descriptive Statistics | | | | | | | | |
| 3 | Credits | 3 | | | | | | | | |
| 4 | Contact Hours | 3-0-0 | | | | | | | | |
| | (L-T-P) | | | | | | | | | |
| | Course Status | DSE | | | | | | | | |
| 5 | Course Objective | 1.To introduce basic statistical concepts, logic and analytical tools and communicatequantitative data verbally, graphically, symbolica numerically. 2.To make students familiar with the concept of Probability and State | | | | | | | | |
| | | display data utilizing various tables, charts, and graphs. | | | | | | | | |
| 6 | Course Outcomes Course Description | CO1: Describe the process and particular steps in designing studies and analyzing data, interpreting and presenting results; and develop presenting quantitative data using appropriate diagrams, tabul summaries. (K2, K5). CO2: Describe the properties of discrete and continuous distributio (K2). CO3: Calculate the measures of central tendency and dispersion describe the method used for analysis, including a discussion of disadvantages, and necessary assumptions. (K2, K3) CO4: Calculate and interpret the correlation between two variable the simple linear regression equation for a set of data and know assumptions behind regression analysis. (K2,K3). CO5: Understand the line of best fit as a tool for summarizin relationship and predicting future observed values, and develop the a formal mathematical argument in the context of probability. (K2, K5) CO6: Develop the skills to interpret the results of statistical analysis. (This is an introductory course in statistics. Students are introduction for a sumption of statistical analysis. | ing skills in ations, and n functions. of data and advantages, s, Calculate w the basic ng a linear bility to use K2, K5). uced to the ences about | | | | | | | |
| | | populations. Included are the study of measures of central ter dispersion, finite probability, statistical inferences from large and sm | | | | | | | | |
| | | linear regression, and correlation. | ~~~ | | | | | | | |
| 8 | Outline syllabus | | CO Mapping | | | | | | | |
| | Unit 1 | Presentation of data | | | | | | | | |
| | А | Classification, tabulation, diagrammatic & graphical representation | CO1 | | | | | | | |
| | В | of groupeddata. Frequency distributions, cumulative frequency distributions | CO1 | | | | | | | |
| | C B | Histogram, Ogives, frequency polygon, Tree and leaf diagram. | C01 | | | | | | | |
| | Unit 2 | Descriptive statistics | CO1 | | | | | | | |
| | | Measures of central tendency – arithmetic mean, median, quartiles, | | | | | | | | |
| | Α | mode, harmonicmean, geometric mean. | CO2 | | | | | | | |
| | В | Their properties, merits, and demerits | CO2 | | | | | | | |
| | С | Measures of dispersion, range, quartile deviation, mean deviation, standard deviation, and coefficient of variation. | | | | | | | | |
| | Unit 3 | Moments | CO3 | | | | | | | |
| | А | Moments, Skewness, Measures of skewness: Karl Pearson's coefficient of skewness. | CO3 | | | | | | | |
| | В | Quartile, coefficient of skewness, Measure of skewness based on moments. | CO3 | | | | | | | |



| С | Kurtosis, Measures of Kurtosis. | |
|------------------|---|-----|
| Unit 4 | Bi-variate data analysis | CO4 |
| А | Bivariate data, principles of least squares, fitting of polynomial curves, and fitting of curves reducible to polynomial form. | CO4 |
| В | Correlation: Spearman's rank correlation, Partial and Multiple Correlation (only two independent variables case). | CO4 |
| С | Regression lines. | |
| Unit 5 | Probability | CO5 |
| A | Probability: Introduction, random experiment, outcomes, sample space, events, various definitions of probability, laws of total and compound probability. | CO5 |
| В | Boole's inequality. Conditional probability, independence of events. | CO5 |
| С | Bayes theorem and its applications in real-life problems. | CO6 |
| Mode exami | | |
| Weigh Distrib | \sim (Δ ·/ Δ %· ENE·/ Δ % | |
| Text b | bk/s* 1. Gupta, S.C. and Kapoor, V.K., "Fundamentals of Mathematical Statistics". | |
| Other | 1. Grewal, B.S, "Higher Engineering Mathematics". | |
| Refere | ces 2. Rohatgi, V.K. "Introduction to Probability". | |

| PO | PO | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| CMS102.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | | |
| CMS102.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 1 | 1 | | |
| CMS102.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 1 | 1 | | |
| CMS102.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | | |
| CMS102.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | | |
| CMS102.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 1 | 1 | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | 1.0 | | |



| Scho | ool: SSBSR | Batch: 2024-28 | |
|----------------|--------------------------|--|----------------|
| | gramme: B.Sc. | Academic Year: 2024-25 | |
| (Hor | | | |
| | ich: Data Science | Semester: I | |
| & A | nalytics Course Code | CSE113 | |
| - | | | |
| 2 | Course Title | Programming for Problem Solving | |
| 3 | Credits | 3 | |
| 4 | Contact Hours (L-T-P) | 3-0-0 | |
| | Course Status | OPE | |
| 5 | Course | To understand and demonstrate how to solve logical and scientific | problems using |
| 5 | Objective | programming. | problems using |
| 6 | Course | CO1: Illustrate and explain the basic computer concepts and program | nming |
| | Outcomes | principles of C language (K2, K3, K4). | |
| | | CO2: Apply and practice the logical ability to solve problems. (K2, | |
| | | CO3: Describe how to generate efficient and schematic solutions to | the problems. |
| | | (K1, K2). | |
| | | CO4: Demonstrate the algorithm, Pseudo-code, and flow chart for the second seco | ne given |
| | | problem (K2, K3, K4). CO5: Create and implement logic using Operators and control states | ments |
| | | CO6: Develop a better understanding of basic concepts of C pro | gramming and |
| | | Computer Organization. | 8 |
| 7 | Course | To understand and demonstrate how to solve logical and scien | tific problems |
| | Description | using C programming. | |
| 8 | | | |
| | Unit 1 | | |
| | A | Introduction to Digital Computers, Representation of Algorithm, Flowcharts, Examples. Introduction to Programming: Importance of C, Basic Structure of C Programs, Programming Style, Executing a C Program. | CO1 |
| | D | Constants, Variables, and Data Types: Introduction, Character Set, C Tokens, Keywords, and Identifiers, Constants, Variables, | CO1 |
| | B | Data Types, Declaration of Variables, Assigning Values to | CO1 |
| | С | Variables, Defining Symbolic Constants. Managing Input and Output Operations: Reading a Character, Writing a Character, Formatted Input, Formatted Output. | CO1 |
| | Unit 2 | | |
| | A | Operators and Expressions: Introduction, Arithmetic Operators, Relational Operators, Logical Operators, Assignment Operators, Increment and Decrement Operators, Conditional Operator, Bitwise Operators, Special Operators, Arithmetic Expressions, Evaluation of Expressions, Precedence of Arithmetic Operators, Type Conversions in Expressions, Operator Precedence and Associativity. | CO2 |
| | В | Decision Making and Branching: Introduction, Decision Making with IF Statement, Simple IF Statement, the IFELSE Statement, Nesting of IFELSE Statements, The ELSE IF Ladder, The Switch statement | CO2 |
| | С | Decision Making and Looping: The WHILE Statement, The DO Statement, the FOR Statement, Jumps in LOOPS. | CO2 |
| | Unit 3 | | |
| | А | Arrays: One-dimensional Arrays, Declaration of One- dimensional Arrays, Initialization of One-dimensional Arrays, | CO3 |
| | В | Two-dimensional Arrays, Declaration of Two-dimensional Arrays, Initialization of Two-dimensional Arrays, | CO3 |
| | С | Example programs – Linear search, Binary search, Bubble sort, and Selection sort. Matrix Multiplication, Transpose of a matrix. | CO3 |



| Unit 4 | | |
|---------------------------|--|----------|
| А | Character Arrays and Strings: Declaring and Initializing String Variables, Reading Strings from the Terminal, Writing Strings to Screen, Arithmetic Operations on Characters, | CO4 |
| В | String-handling Functions (strlen(), strcpy(), strcmp(), strcat(), strrev()), Example Programs (with and without using built-in string functions), Two-dimensional character arrays. | CO4,CO5 |
| С | Introduction, Declaring Pointer Variables, Initialization of Pointer variables, accessing a Variable through its Pointer, Pointer Expressions, Pointer Increments and Scale Factor, Pointers and 1-D Arrays. | CO4, CO5 |
| Unit 5 | | |
| А | Elements of User-defined Functions, Definition of Functions, Return Values and their Types, Function Calls, Function Declaration, Category of Functions | CO5 |
| В | No Arguments and no Return Values, Arguments but no Return values, Arguments with Return Values, No Arguments but Returns a Value, Passing Arrays to Functions. | CO5, CO6 |
| С | Recursion - Factorial of an integer, Xn, Finding n th Fibonacci numbers | CO5, CO6 |
| Mode of | Theory | |
| examination | | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | E. Balagurusamy, Programming in ANSI C, 5th Edition, Tata McGraw-Hill Publications. P B Kottur, Computer Concepts and C Programming | |
| Other References | 1.Kerningham Dennis Ritchie, The C programming language (ANSI C version), 2nd Edition, PHI India 2.Jeri R Hanly Elliot B Koffman, Problem-solving and program design in C Person Addison Wesley | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | PO | РО | PO | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| CSE113.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | | | |
| CSE113.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 1 | | | |
| CSE113.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 1 | | | |
| CSE113.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 1 | | | |
| CSE113.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | | | |
| CSE113.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 1 | | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | | | |



| Scho | ool: SSBSR | Batch: 2024-28 | | | | | | | | |
|---------------------|--------------------------|--|---|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2024-25 | | | | | | | | |
| (Ho | | | | | | | | | | |
| | | Semester: I | | | | | | | | |
| & A 1 | nalytics Course Code | VOM103 | | | | | | | | |
| 2 | Course Title | Essential Excel Skills for Business | | | | | | | | |
| | Course Title Credits | 3 | | | | | | | | |
| 3 | | 3 | | | | | | | | |
| 4 | Contact Hours (L-T-P) | 0-0-3 | | | | | | | | |
| | Course Status | SEC | | | | | | | | |
| 5 | Course | 1. To be able to enter, edit, and format data with ease using the | Excel user | | | | | | | |
| | Objective | interface. | | | | | | | | |
| | 5 | 2. To do calculations on data, use formulae and functions. Utilize | functions to | | | | | | | |
| | | automate selections and data searches. | | | | | | | | |
| 6 | | | | | | | | | | |
| 7 | Course Description | CO6: Create, manage, and apply Named Ranges to enhance calculation In offices all throughout the world, spreadsheet software continues the most frequently used programs. A significant tool will be add employability profile after you learn to use this software with assur- day, there are millions of job postings in India alone that mention h abilities. Digital skills contribute to higher income and better of chances. | to be one of ded to your ance. Every aving Excel | | | | | | | |
| 8 | | | | | | | | | | |
| | Unit 1 | Critical Core of Excel and Performing Calculations | | | | | | | | |
| | А | Introduction, Taking Charge of Excel, Navigating and Selecting, View Options, Data Entry, Data Types, Editing and Deleting, Fill Handle, Copy and Paste, Templates. | CO1 | | | | | | | |
| | В | Formulas, Formulas in Context, Functions I: SUM and AUTOSUM. | CO1 | | | | | | | |
| | С | Functions II: AVERAGE, MIN and MAX, Absolute Cell References, Calculations across sheets. | CO1 | | | | | | | |
| | Unit 2 | Formatting and Printing | | | | | | | | |
| | А | Formatting, Borders, Alignment Tools, Format Painter, Number Formats, Styles, and Themes. | CO2 | | | | | | | |
| | В | Managing Rows and Columns, Finding and Replacing, Filtering, Sorting, Conditional Formatting. | CO2 | | | | | | | |
| | С | Print Preview, Orientation, Margins, Scale, Page Breaks, Print Titles, Headers, and Footers | CO2 | | | | | | | |
| | Unit 3 | Charts | | | | | | | | |
| | А | Basic Chart Types: Pie, Column, and Line Charts. | CO3 | | | | | | | |
| | В | Move and Resize Charts, Change Chart Style & Type. | CO3 | | | | | | | |
| | С | Modify Chart Elements. | CO3C | | | | | | | |
| | Unit 4 | Working with Multiple Worksheets & Workbooks | | | | | | | | |
| | А | Multiple Worksheets, 3D Formulas, Linking Workbooks. | CO4 | | | | | | | |
| | В | Consolidating by Position, Consolidating by Category (Reference). | CO4 | | | | | | | |



| С | Combining Text (CONCAT, &), Changing Text Case (UPPER, LOWER, PROPER). | CO4 |
|--------------|--|-----|
| Unit 5 | Named Ranges | |
| А | Extracting Text (LEFT, MID, RIGHT), Finding Text (FIND), | CO5 |
| В | Date Calculations (NOW, TODAY, YEARFRAC). | CO5 |
| С | Introducing Named Ranges, Creating Named Ranges, Managing Named Ranges, Named Ranges in Formulas, Apply Names. | CO6 |
| Mode of | Practical | |
| examination | | |
| Weightage | CA: 25%; CE: 25%; ETE: 50% | |
| Distribution | , , , | |
| Text book/s* | Michael Alexander, Excel® Dashboards & Reports for Dummies, John Wiley & Sons, Inc, ISBN: 978-1-119-07676-6, 2016. | |
| Other | 1. Michael Alexander and Dick Kusleika, Excel 2016 Formulas, | |
| References | John Wiley & Sons, Inc, ISBN: 978-1-119-06786-3, 2016. | |

| PO | РО | РО | РО | PO | PO | РО | РО | РО | РО | РО | РО | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|-----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| VOM103.1 | | 2 | 1 | 2 | | 1 | | 3 | | | 1 | 1 | 1 | |
| VOM103.2 | | 2 | 1 | 2 | | 1 | | 3 | | | 1 | 1 | 1 | |
| VOM103.3 | | 2 | 1 | 2 | | 1 | | 3 | | | 1 | 1 | 1 | |
| VOM103.4 | | 2 | 1 | 2 | | 1 | | 3 | | | 1 | 1 | 1 | |
| VOM103.5 | | 2 | 1 | 2 | | 1 | | 3 | | | 1 | 1 | 1 | |
| VOM103.6 | | 2 | 1 | 2 | | 1 | | 3 | | | 1 | 1 | 1 | |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 1.0 | 1.0 | 1.0 | |



| Sch | ool: SSBSR | Batch: 2024-28 | | | | | | | | | | |
|----------|---|--|--|--|--|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2024-25 | | | | | | | | | | |
| (Ho | | | | | | | | | | | | |
| | nch: Data Science nalytics | Demester: 1 | | | | | | | | | | |
| <u> </u> | Course Code | ARP101 | | | | | | | | | | |
| 2 | Course Title | Communicative English-1 | | | | | | | | | | |
| 3 | Credits | 2 | | | | | | | | | | |
| 4 | Contact Hours | | | | | | | | | | | |
| т | (L-T-P) | 1-0-2 | | | | | | | | | | |
| | Course Status | AEC | | | | | | | | | | |
| 5 | Course Objective | To minimize the linguistic barriers that emerge in varied environments through the use of English. Help students to und accents and standardize their existing English. Guide the studen basic communication skills - listening, speaking, reading, and w uplifting their perception of themselves, giving them self-confiden- positive attitude. | erstand different nts to hone their riting while also | | | | | | | | | |
| 6 | Course Outcomes Course Description | After completion of this course, students will be able to: CO1: Develop a better understanding of advanced grammar grammatically correct sentences CO2: Acquire wide vocabulary and punctuation rules and learn str free communication. CO3: Interpret texts, and pictures and improve both reading an which would help them in their academic as well as professional ca CO4: Comprehend language and improve speaking skills in acader contexts CO5: Develop, share, and maximize new ideas with the concept and the documentation of key critical thoughts articulated toward career based on their potential and availability of opportunities. CO6: Function effectively in multi-disciplinary teams through th team work, Inter-personal relationships, conflict management, quality The course is designed to equip students, who are at a very basic for comprehension, to communicate and work with ease in the | rategies for error- nd writing skills areer nic and social of brainstorming s preparing for a he knowledge of and leadership level of language | | | | | | | | | |
| | Description | comprehension, to communicate and work with ease in the environment. The course begins with basic grammar structure and patterns, leading up to apprehension of oneself through write expression as a first step towards greater employability. | nd pronunciation | | | | | | | | | |
| 8 | | | | | | | | | | | | |
| | Unit 1 | Sentence Structure | CO Mapping | | | | | | | | | |
| | А | Subject Verb Agreement | CO1 | | | | | | | | | |
| | В | Parts of speech | CO1 | | | | | | | | | |
| | С | Writing well-formed sentences | CO1 | | | | | | | | | |
| | Unit 2 | Vocabulary Building & Punctuation | | | | | | | | | | |
| | A | Homonyms/ homophones, Synonyms/Antonyms | CO1, CO2 | | | | | | | | | |
| | В | Punctuation/ Spellings (Prefixes-suffixes/Unjumbled Words) | CO1, CO2 | | | | | | | | | |
| | C | Conjunctions/Compound Sentences | CO1, CO2 | | | | | | | | | |
| | Unit 3 | Writing Skills | | | | | | | | | | |
| | А | Picture Description – Student Group Activity | CO1 | | | | | | | | | |
| | В | Positive Thinking - Dead Poets Society-Full-length feature film - Paragraph Writing inculcating the positive attitude of a learner through the movie SWOT Analysis – Know yourself | CO3 | | | | | | | | | |



| С | Story Completion Exercise –Building positive attitude - The Man from Earth (Watching a Full-length Feature Film) | CO2, CO3 |
|---------------------------|---|----------|
| | Digital Literacy Effective Use of Social Media | CO2, CO3 |
| Unit 4 | Speaking Skill | CO3 |
| А | Self-introduction/Greeting/Meeting people – Self-branding | CO3 |
| В | Describing people and situations - To Sir with Love (Watching a Full-length Feature Film) | CO4 |
| С | Dialogues/conversations (Situation based Role Plays) | CO4 |
| Unit 5 | Professional Skills Career Skills | CO4 |
| А | Exploring Career Opportunities | CO4, CO5 |
| В | Brainstorming Techniques & Models | CO4, CO5 |
| С | Social and Cultural Etiquettes | CO4, CO5 |
| D | Internal Communication | CO4, CO5 |
| Unit 6 | Leadership and Management Skills | |
| А | Managerial Skills | CO4, CO5 |
| В | Entrepreneurial Skills | CO4, CO5 |
| Mode of examination | Class Assignments/Free Speech Exercises / JAM Group Presentations/Problem-Solving Scenarios/GD/Simulations | |
| Weightage Distribution | 60% CA and 40% ETE | |
| Text book/s* | Blum, M. Rosen. How to Build Better Vocabulary. London: Bloomsbury Publication | |
| Other References | Comfort, Jeremy (et.al). Speaking Effectively. Cambridge University Press | |

| РО | РО | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| ARP101.1 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | | | |
| ARP101.2 | 1 | 2 | 3 | 2 | | 1 | 1 | 3 | 1 | | 1 | | | |
| ARP101.3 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | | | |
| ARP101.4 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | | | |
| ARP101.5 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | | | |
| ARP101.6 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | | | |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 1.0 | | | |



| Scho | ool: SSBSR | Batch: 2024-28 | |
|---------------|-------------------------------|--|--------------------|
| | gramme: B.Sc. | Academic Year: 2024-25 | |
| (Hoi | | | |
| | nch: Data Science nalytics | Semester: I | |
| a A 1 | Course Code | VAC103 | |
| 2 | Course Title | | |
| 2 | Course Title Credits | Environment Management | |
| | - | 3-0-0 | |
| 4 | Contact Hours | 5-0-0 | |
| | (L-T-P) Course Status | VAC | |
| 5 | - | | of |
| 5 | Course | 1.Enable students to learn the concepts, principles, and importance environmental science | 01 |
| | Objective | 2.Provide students an insight into various causes of natural resource | e depletion and |
| | | its conservation | e depiction und |
| | | 3.Provide detailed knowledge of causes, effects, and control of diff | erent types of |
| | | environmental pollution and its effect on climate change, global | warming, and |
| | | ozone layer depletion. | |
| | | 4.Provide knowledge of different methods of water conservation | 1 |
| | | 5. Provide and enrich the students with sustainable practices and en | vironmental |
| 6 | Course | management CO1.Develop a better understanding of the principles and scope o | f environmental |
| 0 | Outcomes | science | r environmental |
| | | CO2. Acquire to learn various pollution causes, effects, and control | and solid waste |
| | | management. | |
| | | CO3. Interpret the effect of global warming and ozone layer depleti | |
| | | CO4. Comprehend various types of natural resources and their cons | |
| | | CO5. Develop a better understanding of sustainable practices and | l environmental |
| | | management CO6. Function effectively an overall understanding of various | environmental |
| | | components, their protection, and management. | environmentai |
| 7 | Course | Environmental Science emphasizes various factors as | |
| | Description | 1. Importance and scope of environmental science | |
| | | 2. Natural resource conservation | |
| | | 3. Pollution causes, effects, and control methods | |
| - | | 4. Sustainable and Environmental environment | |
| 8 | TL •4 1 | | |
| | Unit 1 | Natural resource management | |
| | A | Introduction to Natural Resources | CO1/CO6 |
| | B C | Management of Land and Forest Resources Water and Energy resource Management | CO1/CO6 CO1/CO6 |
| | Unit 2 | Environmental Pollution Management | |
| | | Air pollution Control and Water Pollution treatment Methods | CO2/CO6 |
| | AB | Soil and Noise Pollution Management | CO2/CO6 |
| | С | Solid waste management | CO2/CO6 |
| | Unit 3 | Climate Change Mitigation | |
| | A | Concept of Global Warming and the greenhouse effect | CO3/CO6 |
| | B | Ozone layer Depletion and its consequences | CO3/CO6 |
| | C | Climate change, its effect on the ecosystem, and its mitigation. | CO3/CO6 |
| | | Kyoto protocol and IPCC concerns on changing climate. | |
| | Unit 4 | Natural resource conservation and management | |
| | А | Hot spots, Endangered and endemic species of India | CO4/CO6 |



| В | Threats to biodiversity: habitat loss, poaching of wildlife, man- wildlife conflicts, biological invasions | CO4/CO6 |
|--------------|---|---------|
| С | Conservation of biodiversity: In-situ and Ex-situ conservation of biodiversity. | CO4/CO6 |
| Unit 5 | Sustainable practices and environmental management | |
| А | Sustainable development and sustainable consumption | CO4/CO6 |
| В | Environmental Issues and Management in India | CO4/CO6 |
| С | Environmental Management System (EMS) | CO4/CO6 |
| Mode of | Theory | |
| examination | | |
| Weightage | CA:25%; ESE:75% | |
| Distribution | CA:25%; ESE:75% | |
| Text book/s* | 1. Textbook of Environmental Studies for Undergraduate | |
| | Courses by Erach Bharucha, Pub: Orient Blackswan Pvt Ltd | |
| Other | 2. Environmental Science by G. Tyler Miller, JR. and Scott E. | |
| References | Spoolman; Broks/Cole. | |

| РО | PO | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| VAC103.1 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | | | |
| VAC103.2 | 1 | 2 | 3 | 2 | | 1 | 1 | 3 | 1 | | 2 | | | |
| VAC103.3 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | | | |
| VAC103.4 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | | | |
| VAC103.5 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | | | |
| VAC103.6 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | | | |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 2.0 | | | |



| Scho | ool: SSBSR | Batch: 2024-28 | | | | | | | | | |
|------|--------------------|--|-----------------------|--|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2024-25 | | | | | | | | | |
| (Hor | | | | | | | | | | | |
| | ich: Data Science | Semester: I | | | | | | | | | |
| | nalytics | CMS151 | | | | | | | | | |
| 1 | Course Code | | | | | | | | | | |
| 2 | Course Title | Foundation Course in Mathematics Lab | | | | | | | | | |
| 3 | Credits | | | | | | | | | | |
| 4 | Contact Hours | 0-0-2 | | | | | | | | | |
| | (L-T-P) | | | | | | | | | | |
| | Course Status | CC | | | | | | | | | |
| 5 | Course | To familiarise the students with basic concepts of matrices, deter | minants, and | | | | | | | | |
| | Objective | solving the system of linear equations. | | | | | | | | | |
| | | To understand the basic concept of sets theory, coordinate geometry, on number, and vector algebra. | • | | | | | | | | |
| 6 | Course Outcomes | CO1: The main objective of the course is to equip the student to plo graphs and solve the different types of equations by plotting the graph | using | | | | | | | | |
| | | different computer software such as Mathematica /MATI /Scilab/Maxima etc. (K1, K2, K3) | | | | | | | | | |
| | | CO2. After completion of this course, students would be able | | | | | | | | | |
| | | convergence of sequences through plotting, verify the Bolzan theorem | o-weierstrass | | | | | | | | |
| | | through plotting the sequence, Cauchy's root test by plotting <i>n</i> th root | ots, and Ratio | | | | | | | | |
| | | est by plotting the ratio of <i>n</i> th and $(n + 1)$ th term. (K2,K3) | | | | | | | | | |
| | | CO3. Students would be able to plot Complex numbers and their representations, | | | | | | | | | |
| | | Operations like addition, subtraction, Multiplication, Division, Modulus and | | | | | | | | | |
| | | Graphical representation of polar form. (K2,K3,K4) | | | | | | | | | |
| | | CO4 : Student would be able to perform the following task of the difference of the performance of the perfo | | | | | | | | | |
| | | Addition, Multiplication, Inverse, Transpose, Determinant, Rank, Eigenvalues, Characteristic equation, and verification of the Cay | | | | | | | | | |
| | | theorem, Solving the systems of linear equations. (K2,K3,K4) | ley-naminon | | | | | | | | |
| | | CO5: Develop program scripts and functions using the Mathematica /MATLAB /Maple /Scilab/Maxima development environment. (K3,K4,K5) | | | | | | | | | |
| | | CO6 : Write the program for evaluating linear system of equati | ons, ordinary | | | | | | | | |
| | | differential equations in Mathematica /MATLAB /Maple /Scilab/Max (K4,K5,K6). | | | | | | | | | |
| 7 | Course | To familiarise the students with basic concepts of matrices, deter | minants, and | | | | | | | | |
| | Description | solving the system of linear equations. | | | | | | | | | |
| | | To understand the basic concept of sets theory, coordinate geomet | rv. complex | | | | | | | | |
| | | number, and vector algebra. | <i>J</i> ¹ | | | | | | | | |
| 8 | Outline syllabus | | CO Mapping | | | | | | | | |
| | Unit 1 | List of the practicals to be done using Mathematica /MATLAB | mapping | | | | | | | | |
| | | /Maple /Scilab/Maxima etc. | | | | | | | | | |
| | A, B, C | Plotting the graphs of the following functions: | CO1, CO6 | | | | | | | | |
| | | (i) ax | | | | | | | | | |
| | | (ii) [x] (greatest integer function) | | | | | | | | | |
| | | (iii) $x 2n$; $n \in N$ | | | | | | | | | |
| | | (iv) $x 2n-1$; $n \in \mathbb{N}$ | | | | | | | | | |
| | | $(v) 1; n \in N X 2n-1$ | | | | | | | | | |
| | U:4 3 | (vi) 1 ; $n \in N \times 2n$ List of the presticals to be done using Mathematica (MATLAP | | | | | | | | | |
| | Unit 2 | List of the practicals to be done using Mathematica /MATLAB /Maple /Scilab/Maxima etc. | | | | | | | | | |
| | A, B, C | (vii) $\sqrt{ax + b}$, $ ax + b $, $c \pm ax + b $ | CO1, CO2 | | | | | | | | |
| | A, D, C | $ (\mathbf{v}\mathbf{n}) \mathbf{v}\mathbf{a}\mathbf{x} + 0, \mathbf{a}\mathbf{x} + 0 , 0 \perp \mathbf{a}\mathbf{x} + 0 $ | CO1, CO2 | | | | | | | | |



| | (viii) $ X $, sin (1, x sin 1, eX, e-X for x \neq 0.) () X X X | |
|--------------|---|-----------|
| | (ix) e ax+b, log(ax+b), 1, sin(ax+b), cos(ax+b), | |
| | $ \sin(ax+b) , \cos(ax+b) . ax+b$ | |
| | Observe and discuss the effect of changes in the real constants a | |
| | and b on the graphs. | |
| Unit 3 | List of the practicals to be done using Mathematica /MATLAB | |
| | /Maple /Scilab/Maxima etc. | |
| A, B, C | By plotting the graph find the solution of the equation | CO1, CO2, |
| | x = ex, x2 + 1 = ex, 1 - x2 = ex, x = log10(x), cos(x), etc | CO6 |
| Unit 4 | List of the practicals to be done using Mathematica /MATLAB | |
| | /Maple /Scilab/Maxima etc. | |
| A, B, C | Plotting the graphs of polynomial of degree 2,3, 4 and 5, and their first and second derivatives. | CO2, CO3, |
| | first and second derivatives. | CO4 |
| Unit 5 | List of the practicals to be done using Mathematica /MATLAB | |
| ont 5 | /Maple /Scilab/Maxima etc. | |
| A, B, C | 1. Sketching parametric curves, e.g., Trochoid, Cycloid, Epicycloid | CO4, CO5, |
| | and Hypocycloid etc. | CO6 |
| | 2. Tracing of conic in Cartesian coordinates. | |
| | 3. Graph of circular and hyperbolic functions. | |
| | Obtaining surface of revolution of curves. | |
| Mode of | Practical+Viva | |
| examination | | |
| Weightage | | |
| Distribution | CA:25%; CE:25%; ESE:50% | |
| | 1 MATLAP Differential and Integral Calculus Arrage Creation | |
| Text book/s* | 1. MAT LAB Differential and Integral Calculus, Apress Grayson | |
| | Street Suite 204 Berkely, CA United States | |
| Other | 1. Solving Applied Mathematical Problems with MATLAB, CRC | |
| References | Press. | |

| РО | PO | PO | РО | PO | PO | PO | PO | PO | PO | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| CMS151.1 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | 1 | | |
| CMS151.2 | 1 | 2 | 3 | 2 | | 1 | 1 | 3 | 1 | | 1 | 1 | | |
| CMS151.3 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | 1 | | |
| CMS151.4 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | 1 | | |
| CMS151.5 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | 1 | | |
| CMS151.6 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | 1 | | |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 1.0 | 1.0 | | |



| | ool: SSBSR | Batch: 2024-28 | |
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| | gramme: B.Sc. | Academic Year: 2024-25 | |
| (Ho | | | |
| | nch: Data Science & | & Semester: I | |
| | lytics | CSD112 | |
| 1 | Course Code | CSP113 | |
| 2 | Course Title | Programming for Problem Solving Lab | |
| 3 | Credits | 1 | |
| 4 | Contact Hours (L-T-P) | 0-0-2 | |
| | Course Status | OPE | |
| 5 | Course Objective | To understand and demonstrate how to solve logical and scien using programming C. | tific problems |
| 6 | Course | CO1: How to read, understand and trace the execution of prog | rams written |
| | Outcomes | in C language. (K2, K3, K4). | |
| | | CO2: Apply c programming knowledge to convert the algorith | m into the |
| | | program in C (K2, K3, K4). | |
| | | CO3: Maximize the knowledge of Array and String concepts of | of C |
| | | programming language (K1, K2). | (110 TT) |
| | | CO4: Demonstrate the concept of function, pointers, and struct | ture. $(K3, K4)$ |
| | | K5 | |
| | | CO5: Develop the uses of computers in the engineering industry $V_{\rm CO}$ | ry. (K4, K5, |
| | | K6) CO6: Discuss the more advanced features of the C language (F | (3 K4 K6) |
| 7 | Course | To understand and demonstrate how to solve logical an | d scientific |
| , | Description | problems using programming C. | la selentine |
| 8 | Outline syllabus | proceeding programming of | CO |
| 0 | - | | Mapping |
| | Unit 1 | Lab. Experiment 1: | |
| | A, B, C | Write a c program to swap two numbers with a temporary | CO1, CO2 |
| | | variable. Write a c program to swap two numbers without a | |
| | | temporary variable. | |
| | Unit 2 | Lab. Experiment 2: | |
| | A, B, C | Write a c Program to Add Two Integers. Write a program to | CO2, CO3 |
| | | check given year is leap year. | |
| | Unit 3 | Lab. Experiment 3: | |
| | A, B, C | Write a c program to calculate the average using arrays. Write a c program to find the largest element of the array. | CO3, CO4 |
| | Unit 4 | Lab. Experiment 4: Write a function to calculate the factorial of a number. Write | CO1 CO5 |
| | A, B, C | a c program to store information about students using the | CO4, CO5, CO6 |
| | Unit 5 | structure. Lab. Experiment 5: | |
| | | Write a c program to store information of a student using | CO5 CO(|
| | A, B, C | union. Write a c program to swap two values using pointers. | CO5, CO6 |
| | Mode of | Practical+Viva | |
| | examination | | |
| | Weightage | | |
| | Distribution | CA:25%; CE:25%; ESE:50% | |
| | Text book/s* | 1. Yashavant Kanetkar, "Let Us C", BPB. | |
| | Other | 1. Byron Gottfried, "Programming with C", TMH. | |
| | References | 2.R. G. Dromey, "How to Solve It by Computer", Pearson. | |
| | References | E.R. G. Diomey, How to Solve It by Computer, i carson. | 1 |



| РО | PO | PO | PO | PO | PO | PO | PO | PO | PO | РО | PO | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| CSP113.1 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | 1 | | |
| CSP113.2 | 1 | 2 | 3 | 2 | | 1 | 1 | 3 | 1 | | 2 | 1 | | |
| CSP113.3 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | 1 | | |
| CSP113.4 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | 1 | | |
| CSP113.5 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | 1 | | |
| CSP113.6 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | 1 | | |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 2.0 | 1.0 | | |



| | ool: SSBSR | Batch: 2024-28 | |
|--------------|----------------------------|--|--|
| | gramme: B.Sc. | Academic Year: 2024-25 | |
| <u>(Hor</u> | | | |
| Brar Anal | 1ch: Data Science & | Semester: II | |
| 1 1 | Course Code | CMS131 | |
| 2 | Course Title | | |
| 3 | Credits | Matrix Analysis and Linear Algebra | |
| | | 4 | |
| 4 | Contact Hours | 4-0-0 | |
| | (L-T-P) | | |
| | Course Status | CC | |
| 5 | Course | 1. To familiarize the students with basic concepts of matrices and t | heır |
| | Objective | application in different prospects. | |
| | | 2. To understand the basic concept of linear algebra and inner produ | - |
| 6 | Course | CO1: Describe the concept of the algebra of matrices and eler | |
| | Outcomes | operations and calculate the rank of the matrix and analyze the con | sistency of a |
| | | linear system. (K1, K2, K3) | |
| | | CO2: Explain the concept of Eigenvalues and Eigenvectors; diagonalization of matrices and quadratic & bilinear forms. (K1, K2 | |
| | | CO3: Discuss the basic of Vector spaces. (K2, K3, K4) | , K3) |
| | | CO4: Describe and use the linear transformation and evaluate nulli | ty and kernel. |
| | | (K2, K3, K4) | •) ••••• |
| | | CO5: Explain the range and kernel and the basic introduction of I | nner product |
| | | spaces and orthogonal and orthonormal vectors. (K4, K5) | _ |
| | | CO6: Describe the application of rank, Eigenvalues, Eigenvectors | , and Gram- |
| | | Schmidt orthogonalization. (K4, K5, K6) | |
| 7 | Course | This course introduces the basic algebra of matrices, and their a | |
| | Description | vector space, Linear transformation and its properties, and matrix re | presentation |
| | | of a linear transformation. | <u> </u> |
| 8 | Outline syllabus | | CO Mapping |
| | Unit 1 | Matrix Analysis -I | |
| | А | Course introduction and properties of Matrices, Elementary row | CO 1 |
| | | operations, and Echelon form of a matrix. | CO 1 |
| | | Rank of a Matrix, Normal form of a Matrix, Gauss-Jordan Method: | CO 1 |
| | В | Inverse of a Matrix by elementary operations. | 001 |
| | С | Application of Rank: System of linear homogeneous and non- | |
| | | homogeneous equations, Theorems on the consistency of a system | CO 1, CO 6 |
| | | | <i>,</i> |
| | TL-:4 0 | of linear equations. | , |
| | Unit 2 | Matrix Analysis -II | |
| | Unit 2 A | Matrix Analysis -II Eigenvalues, Eigenvectors, and characteristic equations of a | CO 2, CO 6 |
| | А | Matrix Analysis -II Eigenvalues, Eigenvectors, and characteristic equations of a matrix. | CO 2, CO 6 |
| | A B | Matrix Analysis -IIEigenvalues, Eigenvectors, and characteristic equations of a matrix.Cayley Hamilton theorem and its application, Diagonalization. | |
| | А | Matrix Analysis -IIEigenvalues, Eigenvectors, and characteristic equations of a matrix.Cayley Hamilton theorem and its application, Diagonalization.Quadratic forms, Matrix of quadratic forms, Bilinear forms, Matrix | CO 2, CO 6 |
| | A B C | Matrix Analysis -IIEigenvalues, Eigenvectors, and characteristic equations of a matrix.Cayley Hamilton theorem and its application, Diagonalization.Quadratic forms, Matrix of quadratic forms, Bilinear forms, Matrix of bilinear forms. | CO 2, CO 6 CO 2, CO 5 |
| | A B C Unit 3 | Matrix Analysis -II Eigenvalues, Eigenvectors, and characteristic equations of a matrix. Cayley Hamilton theorem and its application, Diagonalization. Quadratic forms, Matrix of quadratic forms, Bilinear forms, Matrix of bilinear forms. Vector space and Linear Transformations -I | CO 2, CO 6 CO 2, CO 5 |
| | A B C | Matrix Analysis -IIEigenvalues, Eigenvectors, and characteristic equations of a matrix.Cayley Hamilton theorem and its application, Diagonalization.Quadratic forms, Matrix of quadratic forms, Bilinear forms, Matrix of bilinear forms.Vector space and Linear Transformations -I Vector Space, Vector Subspaces and Linear Span, Linear | CO 2, CO 6 CO 2, CO 5 CO 2 |
| | A B C Unit 3 | Matrix Analysis -IIEigenvalues, Eigenvectors, and characteristic equations of a matrix.Cayley Hamilton theorem and its application, Diagonalization.Quadratic forms, Matrix of quadratic forms, Bilinear forms, Matrix of bilinear forms.Vector space and Linear Transformations -IVector Space, Vector Subspaces and Linear Span, Linear Independence, and Linear Dependence, Basic Results on Linear | CO 2, CO 6 CO 2, CO 5 |
| | A B C Unit 3 | Matrix Analysis -IIEigenvalues, Eigenvectors, and characteristic equations of a matrix.Cayley Hamilton theorem and its application, Diagonalization.Quadratic forms, Matrix of quadratic forms, Bilinear forms, Matrix of bilinear forms.Vector space and Linear Transformations -I Vector Space, Vector Subspaces and Linear Span, Linear | CO 2, CO 6 CO 2, CO 5 CO 2 CO 3 |
| | A B C Unit 3 A | Matrix Analysis -IIEigenvalues, Eigenvectors, and characteristic equations of a matrix.Cayley Hamilton theorem and its application, Diagonalization.Quadratic forms, Matrix of quadratic forms, Bilinear forms, Matrix of bilinear forms.Vector space and Linear Transformations -IVector Space, Vector Subspaces and Linear Span, Linear Independence, and Linear Dependence, Basic Results on Linear Independence. | CO 2, CO 6 CO 2, CO 5 CO 2 |
| | A B C Unit 3 A | Matrix Analysis -II Eigenvalues, Eigenvectors, and characteristic equations of a matrix. Cayley Hamilton theorem and its application, Diagonalization. Quadratic forms, Matrix of quadratic forms, Bilinear forms, Matrix of bilinear forms. Vector space and Linear Transformations -I Vector Space, Vector Subspaces and Linear Span, Linear Independence, and Linear Dependence, Basic Results on Linear Independence. Basis of a Finite Dimensional Vector Space, Linear | CO 2, CO 6 CO 2, CO 5 CO 2 CO 3 |



| Unit 4 | Linear Transformations-II | | | | | | | | | | | | |
|---------------------------|---|------|--|--|--|--|--|--|--|--|--|--|--|
| А | Linear operators, Invertible Linear Transformations. | CO 4 | | | | | | | | | | | |
| В | Matrix of a Linear Transformation, Matrix of the sum and product of linear transformations. | CO 4 | | | | | | | | | | | |
| С | Linear transformation of a Quadratic Form and its theorems. | CO 4 | | | | | | | | | | | |
| Unit 5 | Orthogonality | | | | | | | | | | | | |
| А | Inner Product Space (definition and examples), Cauchy- Schwartz inequality. | CO 5 | | | | | | | | | | | |
| В | Orthogonal and orthonormal vectors, Orthogonal and orthonormal bases | CO 5 | | | | | | | | | | | |
| С | Gram-Schmidt Process, Orthogonal, and positive definite matrices. | CO 6 | | | | | | | | | | | |
| Mode of examination | Theory | | | | | | | | | | | | |
| Weightage Distribution | CA:25%; ESE:75% | | | | | | | | | | | | |
| Text book/s* | 1.Strang G, Linear Algebra and its applications, 3rd edition, Thomson. 2.Krishnamurthy V, Mainra V P, Arora J L, An introduction to | | | | | | | | | | | | |
| | Linear Algebra. | | | | | | | | | | | | |
| Other References | 1.Lipshutz S, Lipson M, Linear Algebra, 3 rd edition, Schaum's Outline series. | | | | | | | | | | | | |
| | 2.Kreyszig E, Advanced Engineering Mathematics, John Wiley & sons. | | | | | | | | | | | | |

| РО | PO | PO | РО | PO | PO | PO | PO | PO | РО | РО | PO | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| CMS131.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | | 1 | |
| CMS131.2 | 2 | 2 | 2 | 2 | | 1 | | | | | 1 | | 1 | |
| CMS131.3 | 3 | 3 | 3 | 3 | | 1 | | | | | 1 | | 1 | |
| CMS131.4 | 2 | 2 | 2 | 3 | | 1 | | | | | 1 | | 1 | |
| CMS131.5 | 2 | 3 | 3 | 3 | | 1 | | | | | 1 | | 1 | |
| CMS131.6 | 3 | 2 | 3 | 3 | | 1 | | | | | 1 | | 1 | |
| Average | 2.5 | 2.5 | 2.5 | 2.6 | | 1.0 | | | | | 1.0 | | 1.0 | |



| Scho | ool: SSBSR | Batch: 2024-28 | | | | | | | | | |
|------|---|---|-------------|--|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2024-25 | | | | | | | | | |
| (Hoi | ns.) | | | | | | | | | | |
| | nch: Data Science nalytics | Semester: II | | | | | | | | | |
| 1 | Course Code | CMS132 | | | | | | | | | |
| 2 | Course Title | Mathematical Expectations & Probability Distributions | | | | | | | | | |
| 3 | Credits | 3 | | | | | | | | | |
| 4 | Contact Hours (L-T-P) | 3-0-0 | | | | | | | | | |
| | Course Status | OPE | | | | | | | | | |
| 5 | Course Objective | | | | | | | | | | |
| 6 | inferenceCourse OutcomesCO1: Describe the basic concepts of probability and randomness applications. (K2, K5). CO2: Describe the properties of discrete and continuous random variab CO3: Calculate the measures of central tendency and dispersion of describe the method used for analysis, including a discussion of a disadvantages, and necessary assumptions. (K2, K3) CO4: Calculate and interpret the probability distributions and their a in real life; and limit theorems. (K2, K3). CO5: Monte Carlo simulation of simple probability models, entropy, | | | | | | | | | | |
| 7 | Course | This is an introductory course in probability. Axioms of probability, | conditional | | | | | | | | |
| | Description | probability and independence, Bayes theorem, and probability distribution | ttions. | | | | | | | | |
| 8 | Outline syllabus | | Mapping | | | | | | | | |
| | Unit 1 | Mathematical Expectation Axioms of probability, conditional probability and independence, | 001 | | | | | | | | |
| | A B | Bayes theorem, Random variables: discrete and continuous random variables, probability mass function (p.m.f), probability density function (p.d.f) and cumulative distribution function (c.d.f), Illustrations and | CO1 CO1 | | | | | | | | |
| | С | Properties of random variables. Mathematical Expectation: Expectation of single and bivariate random variables, properties of expectation, conditional expectation, and its properties. Moments and cumulants. Moment generating function, probability generating function. | CO1 | | | | | | | | |
| | Unit 2 | Discrete Random Variable | | | | | | | | | |
| | A | Random variables, distribution function, discrete random variable, expectation, variance | CO2 | | | | | | | | |
| | В | Discrete distributions: Bernoulli and Binomial random variable, Poisson random variable, demerits | CO2 | | | | | | | | |
| | C | Negative binomial random variable, Geometric random variable, and their properties, merits, and demerits | CO2 | | | | | | | | |
| | Unit 3 | Continuous Random Variable | | | | | | | | | |
| | А | Continuous random variable: the expectation of random variable, variance | CO3 | | | | | | | | |
| | В | Continuous distributions: Uniform, Normal, Exponential, Gamma, and Cauchy, computing probabilities by conditioning, moment generating function, their properties, merits, and | CO3 | | | | | | | | |



| | demerits. | |
|--------------|---|----------|
| С | Markov inequality and Chebyshev's inequality. | CO3 |
| Unit 4 | Jointly Distributed Random Variables | CO4 |
| А | Jointly distributed random variables, Independent random variable, the sum of independent random variable | CO4, CO5 |
| В | Central Limit Theorem, conditional distribution with example. | CO4, CO5 |
| С | Joint probability distribution, covariance, correlation coefficient. | |
| Unit 5 | Generation of Random Numbers | |
| А | Generation of random numbers and elements of Monte Carlo simulation. | CO5, CO6 |
| В | Elements of information theory: entropy as a measure of randomness. | CO5,CO6 |
| С | Exploratory data analysis, types of data, frequency tables, descriptive measures, variability measures | CO6 |
| Mode of | Theory | |
| examination | | |
| Weightage | CA:25%; ESE:75% | |
| Distribution | | |
| Text book/s* | 1. Gupta, S.C. and Kapoor, V.K., "Fundamentals of Mathematical Statistics". | |
| Other | 1. Grewal, B.S, "Higher Engineering Mathematics". | |
| References | 2. Rohatgi, V.K. Introduction to Probability. | |

| РО | РО | PO | РО | РО | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| CMS132.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | 1 | |
| CMS132.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 1 | 1 | 1 | |
| CMS132.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 1 | 1 | 1 | |
| CMS132.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | 1 | |
| CMS132.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | 1 | |
| CMS132.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 1 | 1 | 1 | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | 1.0 | 1.0 | |



| Scho | ool: SSBSR | Batch: 2024-28 | | | | | | | | | | |
|------------|-------------------------------|--|------------------|--|--|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2024-25 | | | | | | | | | | |
| (Hor | | | | | | | | | | | | |
| | nch: Data Science nalytics | Semester: II | | | | | | | | | | |
| a A | Course Code | CSE242 | | | | | | | | | | |
| 2 | Course Title | Data Structures | | | | | | | | | | |
| 3 | Credits | | | | | | | | | | | |
| 4 | Contact Hours | | | | | | | | | | | |
| 4 | (L-T-P) | 3-0-0 | | | | | | | | | | |
| | Course Status | CC | | | | | | | | | | |
| 5 | Course | To make students familiar with the data structure & algorithms. The concept of | | | | | | | | | | |
| | Objective | data organizations, data structure operations; analysis of an algorith | hm; Stacks and | | | | | | | | | |
| | | Queues; Linked Lists; Sorting and Hashing; Graph. | | | | | | | | | | |
| 6 | Course Outcomes | CO1: Explain and illustrate the concepts of basic terminologies: e organizations, data structure operations: insertion, deletion, traversa K4) CO2: Describe the analysis of an algorithm, asymptotic; notations trade off $(K1, K2, K3)$ | l, etc. (K2, K3, | | | | | | | | | |
| | | trade-off. (K1, K2, K3) CO3: Describe Linear Search and Binary Search Techniques and complexity analysis. (K2, K3, K4) CO4: Describe ADT Stack and its operations: Algorithms and the analysis, Applications of Stacks; Types of Queue; Algorithms and (K2, K3, K4) CO5: Describe the Singly-linked lists; trees; algorithms and analysis. CO6: Describe and analyze the basic concepts of Sorting and Hash | | | | | | | | | | |
| 7 | Course Description | (K1, K2, K4) This course introduces data structure & algorithms. The concept of data organizations, data structure operations; analysis of an algorithm; Stacks and Queues; Linked Lists; Sorting and Hashing; Graph. | | | | | | | | | | |
| 8 | | | | | | | | | | | | |
| | Unit 1 | | | | | | | | | | | |
| | А | Basic Terminologies: Elementary Data Organizations, | CO1 | | | | | | | | | |
| | В | Data Structure Operations: insertion | CO1 | | | | | | | | | |
| | С | deletion, traversal, etc. | CO1 | | | | | | | | | |
| | Unit 2 | | | | | | | | | | | |
| | А | Analysis of an Algorithm, Asymptotic; | CO2 | | | | | | | | | |
| | В | Notations, Time-Space trade-off. Searching: Linear Search | CO2 | | | | | | | | | |
| | С | Binary Search Techniques and their complexity analysis. | CO2 | | | | | | | | | |
| | Unit 3 | | | | | | | | | | | |
| | A | Stacks and Queues: ADT Stack and its operations: Algorithms and their complexity analysis, | CO3 | | | | | | | | | |
| | В | Applications of Stacks: Expression Conversion and evaluation – corresponding algorithms and complexity analysis. | CO3 | | | | | | | | | |
| | С | ADT queue, Types of Queue: Simple Queue, Circular Queue, Priority Queue; Operations on each type of Queues: Algorithms and their analysis. | CO3 | | | | | | | | | |
| | Unit 4 | | | | | | | | | | | |
| | A | Linked Lists: Singly linked lists: Representation in memory, Algorithms of several operations: Traversing, Searching, Insertion into, Deletion from the linked list; | CO4 | | | | | | | | | |
| | В | Linked representation of Stack and Queue, Header nodes, | CO4 | | | | | | | | | |



| | Doubly linked list: operations on it and algorithmic analysis; Circular Linked Lists: all operations their algorithms and the complexity analysis. | |
|---------------------------|---|--|
| С | Trees: Basic Tree Terminologies, Different types of Trees: Binary Tree, Threaded Binary Tree, Binary Search Tree, AVL Tree; Tree operations on each of the trees and their algorithms with complexity analysis. Applications of Binary Trees. B Tree, B+ Tree: definitions, algorithms, and analysis. | CO5 |
| Unit 5 | Sorting and Hashing: Objective and properties of different sorting algorithms: Selection Sort, Bubble Sort, Insertion Sort, Quick Sort, Merge Sort, Heap Sort; | CO5 |
| А | Performance and Comparison among all the methods, Hashing. | CO5 |
| В | Graph: Basic Terminologies and Representations, Graph search and traversal algorithms, and complexity analysis. | CO6 |
| С | Basic Terminologies: Elementary Data Organizations. | CO6 |
| Mode of examination | Theory | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | 1.Fundamentals of Data Structures", Illustrated Edition by Ellis Horowitz, Sartaj Sahni, Computer Science Press. | |
| Other References | 1.Algorithms, Data Structures, and Problem-Solving with C++", Illustrated Edition by Mark Allen Weiss, Addison-Wesley Publishing Company. 2.How to Solve it by Computer", 2nd Impression by R. G. | |
| | Unit 5 A B C Mode of examination Weightage Distribution Text book/s* Other | Circular Linked Lists: all operations their algorithms and the complexity analysis.CTrees: Basic Tree Terminologies, Different types of Trees: Binary Tree, Threaded Binary Tree, Binary Search Tree, AVL Tree; Tree operations on each of the trees and their algorithms with complexity analysis. Applications of Binary Trees. B Tree, B+ Tree: definitions, algorithms, and analysis.Unit 5Sorting and Hashing: Objective and properties of different sorting algorithms: Selection Sort, Bubble Sort, Insertion Sort, Quick Sort, Merge Sort, Heap Sort;APerformance and Comparison among all the methods, Hashing.BGraph: Basic Terminologies and Representations, Graph search and traversal algorithms, and complexity analysis.CBasic Terminologies: Elementary Data Organizations.Mode of examinationTheoryWeightage DistributionCA:25%; ESE:75%Other References1.Fundamentals of Data Structures", Illustrated Edition by Ellis Horowitz, Sartaj Sahni, Computer Science Press.Other References1.Algorithms, Data Structures, and Problem-Solving with C++", Illustrated Edition by Mark Allen Weiss, Addison-Wesley Publishing Company. |

| РО | РО | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| CSE242.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | | | |
| CSE242.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 1 | | | |
| CSE242.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 1 | | | |
| CSE242.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 1 | | | |
| CSE242.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | | | |
| CSE242.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 1 | | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | | | |



| n | ool: SSBSR | Batch: 2024-28 | |
|----------|--|--|---|
| | gramme: B.Sc. | Academic Year: 2024-25 | |
| (Ho | | | |
| | nch: Data Science | Semester: II | |
| α A 1 | nalytics Course Code | VOM104 | |
| | | | |
| 2 | Course Title | Advanced Excel Skills for Business | |
| 3 | Credits | 3 | |
| 4 | Contact Hours | 0-0-6 | |
| | (L-T-P) | | |
| | Course Status | SEC | |
| 5 | Course | 1. To work through challenges that are all too common that we enco | ounter every |
| | Objective | day. | |
| | | 2. To learn to confidently operate this Excel means adding a highly va | aluable asset |
| | | to the employability portfolio. | |
| 6 | Course | CO1: How to use functions like COUNTIFS to extract information f | rom data, as |
| | Outcomes | well as generate graphical and table representations of it. CO2: Illustrate pivot tables and gain skills to create interactive dash | boards with |
| | | pivot charts and slicers. | iboards with |
| | | CO3: Apply data validation through conditional logic and conditional | format. |
| | | CO4: Analyze functions like CHOOSE, VLOOKUP, INDEX, MATC dynamic lookups to find and display data from several sources. | H, and other |
| | | CO5: Evaluate errors, trace precedents and dependents, and resc | olve circular |
| | | references. | |
| | ~ | CO6: Create protected worksheets and workbooks. | 2.1 |
| 7 | Course | In offices throughout the world, spreadsheet software continues to b | |
| | Description | most frequently used programs. A significant tool will be add | |
| | | employability profile after you learn to use this software with assur day, there are millions of job postings in India alone that mention h | - |
| | | abilities. Digital skills contribute to higher income and better of | U U |
| | | chances. | imployment |
| 8 | | | |
| | | | |
| | Unit 1 | Summarizing Data and Tables | |
| | Unit 1 A | Summarizing Data and Tables COUNT functions, Counting with Criteria (COUNTIFS), Adding | |
| | Unit 1 A | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIES) Sparklines Advanced Charting Trendlines | CO1 |
| | A | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIES) Sparklines Advanced Charting Trendlines | CO1 CO1 |
| | AB | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables | CO1 |
| | A B C | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling | |
| | A B C Unit 2 | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers | CO1 CO1 |
| | A B C Unit 2 A | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table | CO1 CO1 CO2 |
| | A B C Unit 2 A B | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table Value Field Settings, Sorting and Filtering a Pivot Table | CO1 CO1 CO2 CO2 |
| | A B C Unit 2 A B C | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table Value Field Settings, Sorting and Filtering a Pivot Table Reporting Filter Pages, Pivoting Charts, Pivoting Slicers | CO1 CO1 CO2 |
| | A B C Unit 2 A B C Unit 3 | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table Value Field Settings, Sorting and Filtering a Pivot Table Reporting Filter Pages, Pivoting Charts, Pivoting Slicers Data Validation and Conditional Logic | CO1 CO1 CO2 CO2 |
| | A B C Unit 2 A B C | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table Value Field Settings, Sorting and Filtering a Pivot Table Reporting Filter Pages, Pivoting Charts, Pivoting Slicers Data Validation and Conditional Logic Data Validation, Creating Drop-down Lists, Using Formulas in | CO1 CO1 CO2 CO2 |
| | A B C Unit 2 A B C Unit 3 A | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table Value Field Settings, Sorting and Filtering a Pivot Table Reporting Filter Pages, Pivoting Charts, Pivoting Slicers Data Validation and Conditional Logic Data Validation, Creating Drop-down Lists, Using Formulas in Data Validation | CO1 CO1 CO2 CO2 CO2 CO3 |
| | A B C Unit 2 A B C Unit 3 A B | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table Value Field Settings, Sorting and Filtering a Pivot Table Reporting Filter Pages, Pivoting Charts, Pivoting Slicers Data Validation and Conditional Logic Data Validation, Creating Drop-down Lists, Using Formulas in Data Validation Working with Data Validation, Advanced Conditional Formatting | CO1 CO1 CO2 CO2 CO2 CO2 |
| | A B C Unit 2 A B C Unit 3 A | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table Value Field Settings, Sorting and Filtering a Pivot Table Reporting Filter Pages, Pivoting Charts, Pivoting Slicers Data Validation and Conditional Logic Data Validation, Creating Drop-down Lists, Using Formulas in Data Validation Working with Data Validation, Advanced Conditional Formatting Logical Functions I: IF, Logical Functions II: AND, OR, | CO1 CO1 CO2 CO2 CO2 CO3 CO3 |
| | A B C Unit 2 A B C Unit 3 A B | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table Value Field Settings, Sorting and Filtering a Pivot Table Reporting Filter Pages, Pivoting Charts, Pivoting Slicers Data Validation and Conditional Logic Data Validation, Creating Drop-down Lists, Using Formulas in Data Validation Working with Data Validation, Advanced Conditional Formatting Logical Functions I: IF, Logical Functions II: AND, OR, Combining Logical Functions I: IF, AND, OR, Combining Logical | CO1 CO1 CO2 CO2 CO2 CO3 |
| | A B C Unit 2 A B C Unit 3 A B | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table Value Field Settings, Sorting and Filtering a Pivot Table Reporting Filter Pages, Pivoting Charts, Pivoting Slicers Data Validation and Conditional Logic Data Validation, Creating Drop-down Lists, Using Formulas in Data Validation Working with Data Validation, Advanced Conditional Formatting Logical Functions I: IF, Logical Functions II: AND, OR, | CO1 CO1 CO2 CO2 CO2 CO3 CO3 |
| | A B C Unit 2 A B C Unit 3 A B C | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table Value Field Settings, Sorting and Filtering a Pivot Table Reporting Filter Pages, Pivoting Charts, Pivoting Slicers Data Validation and Conditional Logic Data Validation Working with Data Validation, Advanced Conditional Formatting Logical Functions I: IF, Logical Functions II: AND, OR, Combining Logical Functions II: Nested Ifs, Handling Errors: IFERROR, IFNA | CO1 CO1 CO2 CO2 CO2 CO2 CO3 CO3 |
| | A B C Unit 2 A B C Unit 3 A B C Unit 4 | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table Value Field Settings, Sorting and Filtering a Pivot Table Reporting Filter Pages, Pivoting Charts, Pivoting Slicers Data Validation and Conditional Logic Data Validation, Creating Drop-down Lists, Using Formulas in Data Validation Working with Data Validation, Advanced Conditional Formatting Logical Functions I: IF, Logical Functions II: AND, OR, Combining Logical Functions II: Nested Ifs, Handling Errors: IFERROR, IFNA Automating Lookups Introduction to Lookups: CHOOSE | CO1 CO1 CO2 CO2 CO2 CO3 CO3 CO3 CO3 |
| | A B C Unit 2 A B C Unit 3 A B C Unit 3 A C Unit 4 A | COUNT functions, Counting with Criteria (COUNTIFS), Adding with Criteria (SUMIFS), Sparklines, Advanced Charting, Trendlines. Creating and Formatting Tables, Working with Tables, Sorting and Filtering in Tables Automation with Tables, Converting to Range, and Subtotaling Pivot Tables, Charts, and Slicers Creating and Modifying a Pivot Table Value Field Settings, Sorting and Filtering a Pivot Table Reporting Filter Pages, Pivoting Charts, Pivoting Slicers Data Validation and Conditional Logic Data Validation, Creating Drop-down Lists, Using Formulas in Data Validation Working with Data Validation, Advanced Conditional Formatting Logical Functions I: IF, Logical Functions II: AND, OR, Combining Logical Functions II: Nested Ifs, Handling Errors: IFERROR, IFNA Automating Lookups | CO1 CO1 CO2 CO2 CO2 CO3 CO3 CO3 |



| Unit 5 | Formula Auditing and Protection | |
|--------------|--|-----|
| | Error Checking, Formula Calculation Options, Trace Precedents and Dependents | CO5 |
| В | Evaluate Formula, Watch Window | CO5 |
| С | Protecting Workbooks and Worksheets | CO6 |
| Mode of | Practical Based | |
| examination | | |
| Weightage | CA. 250/. ETE, 500/ | |
| Distribution | CA: 25%; ETE: 50% | |
| Text book/s* | Michael Alexander, Excel® Dashboards & Reports for Dummies, John Wiley & Sons, Inc, ISBN: 978-1-119-07676-6, 2016. | |
| Other | 1. Michael Alexander and Dick Kusleika, Excel 2016 Formulas, | |
| References | John Wiley & Sons, Inc, ISBN: 978-1-119-06786-3, 2016. | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | РО | РО | PO | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|-----|----|----|----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| VOM104.1 | | 2 | 1 | 2 | | 1 | | 3 | | | | | 1 | 1 |
| VOM104.2 | | 2 | 1 | 2 | | 1 | | 3 | | | | | 1 | 1 |
| VOM104.3 | | 2 | 1 | 2 | | 1 | | 3 | | | | | 1 | 1 |
| VOM104.4 | | 2 | 1 | 2 | | 1 | | 3 | | | | | 1 | 1 |
| VOM104.5 | | 2 | 1 | 2 | | 1 | | 3 | | | | | 1 | 1 |
| VOM104.6 | | 2 | 1 | 2 | | 1 | | 3 | | | | | 1 | 1 |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | | | 1.0 | 1.0 |



| Sch | ool: SSBSR | Batch: 2024-28 | |
|-----|--|--|--|
| | gramme: B.Sc. | Academic Year: 2024-25 | |
| (Ho | | | |
| | nch: Data Science | Semester: II | |
| | analytics | 4.0.0102 | |
| 1 | Course Code | ARP102 | |
| 2 | Course Title | Communicative English -2 | |
| 3 | Credits | 2 | |
| 4 | Contact Hours | 1-0-2 | |
| | (L-T-P) | | |
| | Course Status | AEC | |
| 5 | Course | To Develop LSRW skills through audio-visual language acquiren | |
| | Objective | writing, advanced speech et al and MTI Reduction with the aid of cer | tain tools like |
| | ~ | texts, movies, and long and short essays. | |
| 6 | Course | After completion of this course, students will be able to: | T . |
| | Outcomes | CO1: Acquire Vision, Goals, and Strategies through Audio-visual Lan | |
| | | CO2: Synthesize complex concepts and present them in creative writin CO3:Develop MTI Reduction/Neutral Accent through Classroom | |
| | | Practice | Sessions a |
| | | CO4: Determine their role in achieving team success by defining | strategies for |
| | | effective communication with different people | strategies for |
| | | CO5: Realize their potential as human beings and conduct themselve | es properly in |
| | | the ways of the world. | 1 1 5 |
| | | CO6 Acquire satisfactory competency in the use of Quantitative | aptitude and |
| | | | |
| | | Logical Reasoning | |
| 7 | Course | The course takes the learnings from the previous semester to an adva | |
| 7 | Course Description | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of | f audio-visual |
| 7 | | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced leve | f audio-visual el of writing, |
| 7 | | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced leve reading, listening, and speaking abilities, while also reducing the usa | f audio-visual el of writing, |
| - | | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced leve | f audio-visual el of writing, |
| 7 8 | Description | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lev- reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. | f audio-visual el of writing, |
| - | | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lev reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual | f audio-visual el of writing, |
| - | Description Unit 1 | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lev- reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts | f audio-visual el of writing, ge of L1 to a |
| - | Description Unit 1 A | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced leve reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life | f audio-visual el of writing, ge of L1 to a CO1 |
| - | Description Unit 1 A B | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lever reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles | f audio-visual el of writing, ge of L1 to a |
| - | Description Unit 1 A | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lever reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action | f audio-visual el of writing, ge of L1 to a CO1 |
| - | Description Unit 1 A B C | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced leve reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action Plans in Life | f audio-visual el of writing, ge of L1 to a CO1 CO1 CO1 |
| - | Description Unit 1 A B C Unit 2 | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lever reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action Plans in Life Creative Writing | f audio-visual el of writing, ge of L1 to a CO1 CO1 CO1 CO2 |
| - | Description Unit 1 A B C Unit 2 A | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lev- reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action Plans in Life Creative Writing Story Reconstruction - Positive Thinking | f audio-visual el of writing, ge of L1 to a CO1 CO1 CO1 CO2 CO2 |
| - | Description Description Unit 1 A B C Unit 2 A B B | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lever reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action Plans in Life Creative Writing Story Reconstruction - Positive Thinking Theme-based Story Writing - Positive attitude | f audio-visual el of writing, ge of L1 to a CO1 CO1 CO1 CO2 |
| - | Description Description Unit 1 A B C Unit 2 A B C C | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lever reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action Plans in Life Creative Writing Story Reconstruction - Positive Thinking Theme-based Story Writing - Positive attitude Learning Diary Learning Log – Self-introspection | f audio-visual el of writing, ge of L1 to a CO1 CO1 CO2 CO2 CO2 |
| - | Description Unit 1 A B C Unit 2 A B C Unit 3 | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lev- reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action Plans in Life Creative Writing Story Reconstruction - Positive Thinking Theme-based Story Writing - Positive attitude Learning Diary Learning Log – Self-introspection Writing Skills 1 | f audio-visual el of writing, ge of L1 to a CO1 CO1 CO2 CO2 CO2 CO2 |
| - | Description Description Unit 1 A B C Unit 2 A B C Unit 3 A | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lever reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action Plans in Life Creative Writing Story Reconstruction - Positive Thinking Theme-based Story Writing - Positive attitude Learning Diary Learning Log – Self-introspection Writing Skills 1 Precis | f audio-visual el of writing, ge of L1 to a CO1 CO1 CO2 CO2 CO2 CO2 CO3 CO3 |
| - | Description Description Unit 1 A B C Unit 2 A B C Unit 3 A B C | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lev reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action Plans in Life Creative Writing Story Reconstruction - Positive Thinking Theme-based Story Writing - Positive attitude Learning Diary Learning Log – Self-introspection Writing Skills 1 Precis Paraphrasing | f audio-visual el of writing, ge of L1 to a CO1 CO1 CO2 CO2 CO2 CO2 |
| - | Description Unit 1 A B C Unit 2 A B C Unit 3 A B C C | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lev reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action Plans in Life Creative Writing Story Reconstruction - Positive Thinking Theme-based Story Writing - Positive attitude Learning Diary Learning Log – Self-introspection Writing Skills 1 Precis Paraphrasing Essays (Simple essays) | f audio-visual el of writing, ge of L1 to a CO1 CO1 CO2 CO2 CO2 CO2 CO3 CO3 |
| - | Description Description Unit 1 A B C Unit 2 A B C Unit 3 A B C | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lev reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action Plans in Life Creative Writing Story Reconstruction - Positive Thinking Theme-based Story Writing - Positive attitude Learning Diary Learning Log – Self-introspection Writing Skills 1 Precis Paraphrasing Essays (Simple essays) MTI Reduction/Neutral Accent through Classroom Sessions & Practice | f audio-visual el of writing, ge of L1 to a CO1 CO1 CO2 CO2 CO2 CO2 CO3 CO3 |
| - | Description Unit 1 A B C Unit 2 A B C Unit 3 A B C C | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lever reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action Plans in Life Creative Writing Story Reconstruction - Positive Thinking Theme-based Story Writing - Positive attitude Learning Diary Learning Log – Self-introspection Writing Skills 1 Precis Paraphrasing Essays (Simple essays) MTI Reduction/Neutral Accent through Classroom Sessions & Practice Vowel, Consonant, sound correction, speech sounds, Monothongs, | f audio-visual el of writing, ge of L1 to a CO1 CO1 CO2 CO2 CO2 CO2 CO3 CO3 CO3 |
| - | Description Description Unit 1 A B C Unit 2 A B C Unit 3 A B C Unit 3 A B C Unit 4 | The course takes the learnings from the previous semester to an adva language learning and self-comprehension through the introduction of aids as language enablers. It also leads learners to an advanced lev reading, listening, and speaking abilities, while also reducing the usa minimum to increase employability chances. Acquiring Vision, Goals, and Strategies through Audio-visual Language Texts Pursuit of Happiness / Goal Setting & Value Proposition in life 12 Angry Men / Ethics & Principles The King's Speech / Mission statement in life strategies & Action Plans in Life Creative Writing Story Reconstruction - Positive Thinking Theme-based Story Writing - Positive attitude Learning Diary Learning Log – Self-introspection Writing Skills 1 Precis Paraphrasing Essays (Simple essays) MTI Reduction/Neutral Accent through Classroom Sessions & Practice | f audio-visual el of writing, ge of L1 to a CO1 CO1 CO2 CO2 CO2 CO2 CO3 CO3 CO3 CO3 CO3 |



| Unit 5 | Gauging MTI Reduction Effectiveness through Free Speech | |
|---------------------------|--|-----|
| А | Jam sessions | CO4 |
| В | Extempore | |
| С | Situation-based Role Play | CO5 |
| Unit 6 | Leadership and Management Skills | |
| А | Innovative Leadership and Design Thinking | CO5 |
| В | Ethics and Integrity | CO5 |
| Unit 7 | Universal Human Values | |
| А | Love & Compassion, Non-Violence & Truth | CO5 |
| В | Righteousness, Peace | CO6 |
| С | Service, Renunciation (Sacrifice) | CO6 |
| Unit 8 | Introduction to Quantitative aptitude & Logical Reasoning | |
| А | Analytical Reasoning & Puzzle Solving | CO6 |
| В | Number Systems and its Application in Solving Problems | CO6 |
| Mode of | Class Assignments/Free Speech Exercises / JAM Group | |
| examination | Presentations/Problem-Solving Scenarios/GD/Simulations | |
| Weightage Distribution | 60% CA and 40% ETE | |
| Text book/s* | Wren, P.C.&Martin H. High English Grammar and Composition, S.Chand& Company Ltd, New Delhi. | |
| Other References | Blum, M. Rosen. How to Build Better Vocabulary. London: Bloomsbury Publication Comfort, Jeremy(et.al). Speaking Effectively. Cambridge University Press. The Luncheon by W. Somerset Maugham - http://mistera.co.nf/files/sm_luncheon.pdf | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | РО | PO | PO | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|-----|-----|-----|----|----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| ARP102.1 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | | | | |
| ARP102.2 | 1 | 2 | 3 | 2 | | 1 | 1 | 3 | 1 | | | | | |
| ARP102.3 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | | | | |
| ARP102.4 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | | | | |
| ARP102.5 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | | | | |
| ARP102.6 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | | | | |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | | | | |



| Scho | ool: SSBSR | Batch: 2024-28 | | | | | | |
|------|--------------------------|--|-------------------------------|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2024-25 | | | | | | |
| (Hor | | | | | | | | |
| Brai | | Semester: II | | | | | | |
| | putational hematics & | | | | | | | |
| | istics | | | | | | | |
| 1 | Course Code | VAC110 | | | | | | |
| 2 | Course Title | Yoga for Holistic Health | | | | | | |
| 3 | Credits | 3 | | | | | | |
| 4 | Contact Hours (L-T-P) | 0-1-4 | | | | | | |
| | Course Status | VAC | | | | | | |
| 5 | Course | To make the students familiar with the different practices of yo | oga, chanting | | | | | |
| | Objective | and meditation techniques and learn the correct teaching skills. | | | | | | |
| 6 | Course Outcomes | CO1: To make the students understand the concept of health and through Yoga CO2 To define the concept and principles of Yoga. CO3: To interpret and understand the breathing practice. CO4: To describe the knowledge about Yoga, its foundations and the breathing practice. | ıd | | | | | |
| | | applications to the aspirants. CO5: To make students aware of Yogic impact on the positive health ar personality development. CO6: The students will learn primary level of Yoga practices, which we groom their personality. | | | | | | |
| 7 | Course Description | | | | | | | |
| 8 | | | | | | | | |
| | Unit 1 | Importance of Health, Wellness through Yoga | | | | | | |
| | A | Meaning, Definition, Aim of Yoga; Concept of health according to WHO and Ayurveda | CO1, CO2, CO4, CO5, CO6 | | | | | |
| | В | Misconception about Yoga, Difference between asana and physical exercise | CO1, CO2, CO4, CO5, CO6 | | | | | |
| | С | Need, Importance of Yoga in health and wellness | CO1, CO2, CO4, CO5, CO6 | | | | | |
| | Unit 2 | Schools of Yoga, Modern and Ancient schools of Yoga existing in India, Yogic diet, Yogic attitudes, Sadhak tatva & Badhak tatva | | | | | | |
| | А | Schools/ Streams of Yoga – Ashtanga Yoga, Bhakti Yoga, Karma | CO3, CO4, | | | | | |
| | | Yoga, Jnana Yoga | CO5, CO6 | | | | | |
| | В | Modern and ancient schools of Yoga existing in India – Natha Sampradaya, Kaivalyadhama, Bihar School of Yoga, Munger, Pragya Yoga (Shantikunj), Iyengar Yoga, Patanjali Yoga Peeth, Ashtanga Vinyasa Yoga | CO3, CO4, CO5, CO6 | | | | | |
| | С | Yoga Ahaara (Yogic diet), Yogic Attitudes – Maitri Karuna, Mudita, | CO3, CO4, | | | | | |
| | | Upeksha, Sadhak Tatva Badhak Tatva (facilitating/helping factors and obstacles in Yoga sadhana) | CO5, CO6 | | | | | |
| | Unit 3 | Beginner level practices – Sukshma Vyayama and Surya Namaskara | | | | | | |
| | А | | CO4, CO5, | | | | | |
| | 1 | Sukshma Vyayama and their benefits for health Part-1 (Bihar | CO6 | | | | | |



| | School of Yoga) Part-1 | |
|---------------------------|--|--------------------|
| В | Sukshma Vyayama & their benefits for health (Swami Dhirendra Brahmachari) Part-1 | CO4, CC CO6 |
| С | Surya Namaskara (Sun Salutation) with mantra chanting (12 steps) & their benefits for health | CO4, CC CO6 |
| Unit 4 | Asana - all categories | |
| А | Standing & Sitting - Tadasana, Vrikshasana, Katichakrasana, Padmasana, Vajrasana, Ushtrasana, Paschimottanasana, Vakrasana | CO4, CC CO6 |
| В | Supine and Prone: Uttanapadasana, Pawanamuktasana, Shalabhasana, Bhujangasana | CO4, CO CO6 |
| С | Balancing and Inverted: Trivikramasana, Sarvangasana, Viparitakarani mudra | CO4, CO CO6 |
| Unit 5 | Pre-practices of Pranayama, Pranayama and Dhyana | |
| A | Kapalabhati, Mukha dhauti, Vibhagiya pranayama (Sectional breathing) | CO1, CO CO5, CO |
| В | Anuloma – Viloma, Bhastrika, Shitali | CO1, CO CO5, CO |
| С | Om Dhyana, Aanapaanasati Dhyana (breath meditation) | CO1, CO CO5, CO |
| Mode of examination | Theory and Practical | |
| Weightage Distribution | CA:60%; ESE:40% | |
| Text book/s* | Sri Ananda: The Complete book of Yoga, Orient Course Backs, Delhi,2003. | |
| Other References | Sri Ananda: The Complete book of Yoga, Orient Course Backs, Delhi,2003. Basavaraddi, I.V. & other: SHATKARMA: A Comprehensive description about Cleansing Process, MDNIY New Delhi, 2009 Joshi, K.S.: Yogic Pranayama, Oriental Paperback, New Delhi, 2009 Dr. Nagendra H R: Pranayama, The Art & Science, Swami VivekanandaYoga Prakashan, Bangalore, 2005. Swami Niranjanananda Saraswati: Asana Pranayama Mudra Bandha, Yoga Publication Trust, Munger Bihar. Joshi, K.S.: Yogic Pranayama, Oriental Paperback, New Delhi, 2009 Swami Kuvalyananda: Pranayama, Kaivalyadhama, Lonavla, 2010 | |



| 8. Swami Rama: Science of Breath, A Practical |
|--|
| Guide, The Himalayan International Institute, |
| Pennselvenia, 1998. |
| 9. Swami Niranjanananda Saraswati: Prana, Pranayama & |
| Pranavidya, Yoga Publications Trust, Munger, Bihar, 2005 |

| PO | PO | PO | РО | PO | PO | PO | PO | PO | PO | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| VAC110.1 | 1 | 3 | 3 | 3 | 2 | 1 | 2 | 3 | 2 | 3 | 2 | 1 | 3 | 3 |
| VAC110.2 | 1 | 2 | 3 | 1 | 3 | 1 | 3 | 2 | 2 | 3 | 1 | 1 | 2 | 3 |
| VAC110.3 | 1 | 1 | 3 | 3 | 3 | 3 | 2 | 3 | 2 | 3 | 2 | 1 | 1 | 3 |
| VAC110.4 | 1 | 2 | 3 | 3 | 1 | 2 | 3 | 2 | 3 | 2 | 1 | 1 | 2 | 3 |
| VAC110.5 | 2 | 2 | 3 | 3 | 1 | 3 | 3 | 2 | 3 | 1 | 2 | 2 | 2 | 3 |
| VAC110.6 | 3 | 3 | 2 | 2 | 3 | 1 | 2 | 3 | 1 | 2 | 3 | 3 | 3 | 2 |
| Average | 1.5 | 2.2 | 2.8 | 2.5 | 2.2 | 1.8 | 2.5 | 2.5 | 2.2 | 2.3 | 1.8 | 1.5 | 2.2 | 2.8 |



| Sch | ool: SSBSR | Batch: 2024-28 | |
|-----------------|---------------------------|--|-------------------|
| | gramme: B.Sc. | Academic Year: 2024-25 | |
| (Ho | | | |
| | nch: Data Science | & Semester: II | |
| Ana 1 | llytics Course Code | CMS171 | |
| 2 | Course Title | | |
| 2 | Course Thie | Matrix Analysis and Linear Algebra Lab | |
| | | 1 | |
| 4 | Contact Hours (L-T-P) | 0-0-2 | |
| | Course Status | CC | |
| 5 | Course | 1. To familiarize the students with the use of MATLAB in Matr | ix analysis. |
| | Objective | 2. To understand the use of MATLAB in Linear Algebra. | - |
| 6 | Course | The student will be able to write a code in Mathematica /MA | ATLAB /Maple |
| | Outcomes | /Scilab/Maxima | r |
| | | CO1: to transform a matrix into echelon form and to find the rank | . (K1, K2, K3) |
| | | CO2: to find the inverse, and eigenvalues & eigenvectors of a | matrix and also |
| | | the solution of a system of equations. (K1, K2, K3) | |
| | | CO3: to verify Cayley-Hamilton theorem. (K2, K3) | |
| | | CO4: to understand Quadratic and Bilinear forms with the help | o of MATLAB. |
| | | (K3, K4, K5) | |
| | | CO5: to apply the concept for vectors linear dependency and inc | lependency and |
| | | also Linear Transformations. (K4, K5, K6) | : |
| | | CO6: to discuss the Gram-Schmidt Process and the concept of e eigenvectors. (K4, K5, K6) | igenvalues and |
| 7 | Course | The course is an introduction to MATLAB in Matrix analysis | sis and Linear |
| | Description | algebra. The primary objective of the course is to develop basic | |
| | 1 | modeling and solve various equations using MATLAB. | |
| 8 | Outline syllabus | | CO Mapping |
| | Unit 1 | | |
| | A, B, C | Algebra of Matrices, Echelon form of a Matrix, | |
| | | Echelon form of a Matrix, Rank of a Matrix. | CO 1 |
| | Unit 2 | | |
| | A, B, C | Gauss-Jordan Method for finding Inverse, System of Equations, | colored colored (|
| | | Eigenvalues, eigenvectors, | CO 2, CO 6 |
| | Unit 3 | | |
| | A, B, C | Matrix of a Quadratic form, | |
| | , , | Matrix of a Bilinear form, | CO 3 |
| | Init 4 | Cayley Hamilton Theorem. | |
| | Unit 4 | Linear dependence and linear independence of vectors, | |
| | A, B, C | Linear Transformation, Inner Product Space | CO 4 |
| | Unit 5 | | |
| | A, B, C | Orthogonal Vectors, | |
| | 1, 2, 0 | Orthonormal Vectors, | CO 5, CO 6 |
| | | Gram-Schmidt Process. | |
| | Mode of | Practical | |
| | examination | | |
| | | | |
| | Weightage Distribution | CA:25%; CE:25%; ESE:50% | |



| ICAT DOOK/S | B.D. Hahn, Essential MATLAB for Scientists and Engineers, John Wiley & Sons, New York, NY. D.R. Hill and D.E. Zitarelli, Linear Algebra Labs with MATLAB, Second edition, Prentice Hall, Upper Saddle River. | |
|-------------|---|--|
| References | R.E. Larson and B.H. Edwards, Elementary Linear Algebra, Third edition, D.C. Heath and Company, Lexington, MA. S.J. Leon, Linear Algebra with Applications, Fifth edition, Prentice Hall, Upper Saddle River. | |

| РО | РО | PO | РО | РО | PO | PSO | PSO | PSO |
|----------|----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| CMS171.1 | | 3 | 2 | 3 | 2 | 3 | 2 | 3 | 2 | 2 | 1 | 1 | 2 | 1 |
| CMS171.2 | | 3 | 3 | 3 | 3 | 2 | 3 | 3 | 2 | 2 | 1 | 1 | 2 | 1 |
| CMS171.3 | | 3 | 2 | 2 | 2 | 2 | 3 | 3 | 2 | 2 | 1 | 1 | 2 | 1 |
| CMS171.4 | | 3 | 2 | 3 | 2 | 3 | 2 | 3 | 3 | 2 | 1 | 1 | 2 | 1 |
| CMS171.5 | | 3 | 2 | 2 | 2 | 2 | 3 | 3 | 2 | 2 | 1 | 1 | 2 | 1 |
| CMS171.6 | | 3 | 2 | 3 | 2 | 3 | 2 | 3 | 3 | 2 | 1 | 1 | 2 | 1 |
| Average | | 3.0 | 2.2 | 2.7 | 2.7 | 2.5 | 2.5 | 3.0 | 2.3 | 2.0 | 1.0 | 1.0 | 2.0 | 1.0 |



| Sch | ool: SSBSR | Batch: 2024-28 | | | | | | |
|------|-------------------------------|---|---|--|--|--|--|--|
| (Hoi | | Academic Year: 2024-25 | | | | | | |
| | nch: Data Science & lytics | Semester: II | | | | | | |
| 1 | Course Code | CSP242 | | | | | | |
| 2 | Course Title | Data Structures Lab | | | | | | |
| 3 | Credits | 1 | | | | | | |
| 4 | Contact Hours(L- T-P) | 0-0-2 | | | | | | |
| | Course Status | CC | | | | | | |
| 5 | Course Objective | To make students familiar with the data structure & algorithms. Th data organizations, data structure operations; analysis of an algorith and Queues; Linked Lists; Sorting and Hashing; Graph. | e concept of m; Stacks | | | | | |
| 6 | Course Outcomes Course | CO1: Explain and illustrate the concepts of basic terminologies data organizations, data structure operations: insertion, deletion, t (K2, K3, K4) CO2: Describe the analysis of an algorithm, asymptotic; notations, trade-off. (K1, K2, K3) CO3: Describe Linear Search and Binary Search Techniques and complexity analysis. (K2, K3, K4) CO4: Describe ADT Stack and its operations: Algorithms and the analysis, Applications of Stacks; Types of Queue; Algorithm analysis. (K2, K3, K4) CO5: Describe the Singly-linked lists; trees; algorithms and analy K6) CO6: Describe and analyze the basic concepts of Sorting and Hast (K1,K2, K4) This course introduces data structure & algorithms. The cond | a time-space explain their ir complexity ns and their ysis. (K2, K3, hing; Graphs. | | | | | |
| | Description | organizations, data structure operations; analysis of an algorithm Queues; Linked Lists; Sorting and Hashing; Graph. | | | | | | |
| 8 | Outline syllabus | | CO Mapping | | | | | |
| | Unit 1 | Lab. Experiment 1: | | | | | | |
| | A, B, C | Problem-based on uses functions to perform the following operations on a singly linked list i) Creation ii) Insertion iii) Deletion iv) Traversal. Problem-based on uses functions to perform the following operations on the doubly linked list i) Creation ii) Insertion iii) Deletion iv) Traversal. | CO1, CO2 | | | | | |
| | Unit 2 | Lab. Experiment 2: | | | | | | |
| | A, B, C | Problem-based on uses functions to perform the following operations on circular linked List i) Creation ii) Insertion iii) Deletion iv) Traversal. Problem-based on implement stack (its operations) using i) Arrays ii) Linked list(Pointers). | CO2, CO3 | | | | | |
| - | Unit 3 | Lab. Experiment 3: | | | | | | |
| | A, B, C | Problem-based on implementing Queue (its operations) using i) Arrays ii) Linked list (Pointers). Problem-based on implementing Circular Queue using arrays. Problem-based on both recursive and nonrecursive functions to perform the following searching operations for a Key value in a given list of integers: a) Linear search b) Binary search. | CO3, CO4 | | | | | |
| | Unit 4 | Lab. Experiment 4: | | | | | | |
| | A, B, C | Problem-based on implements the following sorting i) Bubble sort ii) Selection sort iii) Quick sort. Problem-based on implements the following i) Insertion sort ii) Merge sort iii) Heap sort. Problem-based on implementing all the functions of a dictionary (ADT) using Linked List. | CO4, CO5, CO6 | | | | | |
| | Unit 5 | Lab. Experiment 5: | | | | | | |



| A, B, C | an element into a binary search tree. b) Delete an element from a binary search tree. c) Search for a key element in a binary search tree. Problem-based on to implement the tree traversal methods. Problem-based on performing the following operations: a) Insert an element into an AVL tree. b) Delete an element from an AVL tree. c) Search for a key element in an AVL tree. | | | | | | |
|---------------------------|---|--|--|--|--|--|--|
| Mode of examination | Practical+Viva | | | | | | |
| Weightage Distribution | CA:25%; CE:25%; ESE:50% | | | | | | |
| Text book/s* | 1. Fundamentals of Data Structures", Illustrated Edition by Ellis Horowitz, SartajSahni, Computer Science Press. | | | | | | |
| Other References | Algorithms, Data Structures, and Problem-Solving with C++", Illustrated Edition by Mark Allen Weiss, Addison-Wesley Publishing Company. How to Solve it by Computer", 2nd Impression by R. G. Dromey, Pearson Education. | | | | | | |

| РО | РО | PO | РО | PO | PO | PO | PO | PO | PO | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| CSP242.1 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | | | |
| CSP242.2 | 1 | 2 | 3 | 2 | | 1 | 1 | 3 | 1 | | 1 | | | |
| CSP242.3 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | | | |
| CSP242.4 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | | | |
| CSP242.5 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | | | |
| CSP242.6 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | | | |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 1.0 | | | |



Detailed Syllabus for

DIPLOMA IN

DATA SCIENCE & ANALYTICS



| Scho | ol: SSBSR | Batch: 2024-28 | |
|-----------------|--------------------------|--|--|
| | ramme: B.Sc. | Academic Year: 2025-26 | |
| (Hon | | | |
| | ich: Data Science | Semester: III | |
| & Al | nalytics Course Code | MSM312 | |
| 2 | Course Code | | |
| 2 | | Discrete Mathematics | |
| | Credits | 4 | |
| 4 | Contact Hours (L-T-P) | 3-1-0 | |
| | Course Status | DSE | |
| 5 | Course | This course is aimed to provide an advanced understanding of sets and | d |
| | Objective | propositions, relations and functions, permutation and combination, g groups, and rings. | raphs, |
| 6 | Course Outcomes | CO1: Discuss the concept of sets, un-countably infinite sets, the inclusion and exclusion, multisets, propositions, and conditional pro evaluate normal forms, Mathematical induction. (K2, K3, K4, K5) CO2: Describe the concept functions, the composition of function functions, and discrete properties of binary relations and check t relations. (K3, K6) CO3: Explain the concept of POSET and lattices, Warshall Equivalence relations, and partitions, and evaluate Chains and Generating Functions, Recurrence relations, and discussing line relations with constant coefficient, homogeneous solution, total s solutions by method of Generating function. (K2, K4, K5) CO4: Illustrate the concept permutations and combinations: rule product, write the algorithms for the generation of permutations and (K3, K5, K6) CO5: Discuss the concept graph, sub-graph, Walks, Path and circu | positions, and ons, invertible he closure of 's algorithm, Anti-chains. ar recurrence solutions, and of sum and combination. |
| 7 | Course Description | graphs, disconnected graphs, and components, and evaluate the circuits, distance, diameters, radius, and pendant vertices, rooted an (K1, K2, K5, K6) CO6: Demonstrate an understanding of Algebraic systems, Group Semi-groups, Monoid, Subgroups, Isomorphism, and Automorphism. This course is given a deep knowledge of sets and propositions, r functions, permutation and combination, graphs, groups, and rings. | d binary trees and evaluate (K2, K5) elations and |
| 8 | Outline syllabus | | CO Mapping |
| | Unit 1 | Sets and Propositions | |
| | А | Sets, Un-countably infinite sets, Principle of inclusion and exclusion, multisets, propositions, | CO1 |
| | В | Conditional propositions. Logical connectivity, Propositional, calculus, | CO1, CO2 |
| | С | Universal and existential quantifiers, Normal forms, methods of proofs, Mathematical induction. | CO2 |
| | Unit 2 | Relations and Functions | CO3 |
| | А | Functions, Composition of function, invertible functions, Discrete properties of binary relations, closure of relations | CO3 |
| | B | Warshall's algorithm, Equivalence relations and partitions, Ordered Sets and Lattices: Introduction, Ordered set, | CO3 |
| | С | Hasse diagram of partially ordered set, Consistent enumeration, Isomorphic ordered set, Well ordered set, Lattices, Properties of | |



| | lattices, Bounded lattices, Distributive lattices, and Complemented | |
|---------------------------|---|-----|
| Unit 3 | lattices. Chains, and Anti-chains. Number Theory | CO4 |
| A | Counting: Basic counting principles, factorial notation, Binomial coefficients, Ordered and unordered partitions. | CO4 |
| В | Permutations and combinations: Rule of sum and Product, Permutations, Combination, Algorithms for Generation of Permutations and Combination, | CO4 |
| С | The Pigeonhole principle, Fundamental theorem of arithmetic, Congruence relation, Congruence Equations. | |
| Unit 4 | Recurrence Relations and Algebraic Structures | CO5 |
| А | Discrete Numeric Functions and Generating functions, | CO5 |
| В | Simple Recurrence relation with constant coefficients | CO5 |
| С | Linear recurrence relations without constant coefficients, Asymptotic behavior of functions. | |
| Unit 5 | Algebraic Structures | CO6 |
| А | Algebraic systems, Group, Semi-groups, Monoid, Subgroups. | CO6 |
| В | Cyclic group, Permutation groups, Homomorphism, | CO6 |
| С | Isomorphism and Automorphism of groups. | |
| Mode of examination | Theory | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | 1.Liu C.L. and Mohapatra, D.P., Elements of Discrete Mathematics", SiE edition, TMH, 2008 | |
| Other References | 1.Kenneth H.R.,' Discrete Mathematics and its Applications", Mc graw hill. 2.Biggs N., "Discrete Mathematics", 3rd edition, Oxford University | |

| РО | PO | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MSM312.1 | | 3 | 2 | 2 | | 1 | | | | | 1 | | | |
| MSM312.2 | | 3 | 3 | 2 | | 1 | | | | | 1 | | | |
| MSM312.3 | | 2 | 2 | 3 | | 1 | | | | | 1 | | | |
| MSM312.4 | | 3 | 2 | 2 | | 1 | | | | | 1 | | | |
| MSM312.5 | | 3 | 2 | 2 | | 1 | | | | | 1 | | | |
| MSM312.6 | | 3 | 2 | 3 | | 1 | | | | | 1 | | | |
| Average | | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | | | |



| Scho | ol: SSBSR | Batch: 2024-28 | |
|-------------|--------------------------|--|--|
| | ramme: B.Sc. | Academic Year: 2025-26 | |
| (Hon | | | |
| | ch: Data Science | Semester: III | |
| a Al | nalytics Course Code | BDA215 | |
| 2 | Course Title | Operations Research | |
| | | operations Research | |
| 3 | Credits | 5 | |
| 4 | Contact Hours (L-T-P) | 3-0-0 | |
| 0 | Course Status | OPE | |
| 5 | Course Objective | To familiarize the students with basic concepts of optimiz classification of optimization problems. To understand the basic concept of Formulation simplex method with upper bounds. | |
| 6 | Course Outcomes | Students will be able to: CO1: Explain the fundamental knowledge of Linear Programming Duality problems. (K1, K2, K3). CO2: Use classical optimization techniques and numerical optimization. (K2, K3, K4). CO3: Describe the basics of different NLPP and KKT conditions. (k3) CO4: Enumerate fundamentals of Integer programming technique different techniques to solve various optimization problems a engineering areas. (K2, K3, K4). CO5: Students will understand the concept of LPP and NLPP and w solve some real-life problems using optimization techniques. (K3, K4) CO6: Explain the fundamental knowledge of Linear Programming a Programming problems. (K4, K5, K6). | methods of , K4). e and apply arising from ill be able to . K5) |
| 7 | Course Description | This course is an introduction to the basic understanding of with app scope of O.R. Formulation of linear programming problems and th methods to solve them will be discussed. Duality in LPP will be int introduction to NLPP and some solving methods will be covered. KKT Conditions, Unconstrained and constrained optimization techni discussed. | nen different roduced. An At the end |
| 8 | Outline syllabus | | CO |
| | TT | Introduction to LDD. Combined Mathed and Simple Mathed | Mapping |
| | Unit 1 A | Introduction to LPP, Graphical Method, and Simplex Method Introduction to Optimization, Assumptions & Mathematical Modeling of LPP, Graphical Solution of L.P.P., Graphical Solution of LPP-I, Graphical Solution of LPP- II. | CO1 |
| | В | Solution of L.P.P. by Simplex method, Revised Simplex Method, Introduction of Big M method, Algorithm of BIG-M method. | CO1 |
| | С | Problems on BIG-M Method, Two Phase Method: Introduction and Two-Phase Method: Problem Solution. | CO1 |
| | Unit 2 | Duality Theory and Integer Programming | |
| | A | Special Cases of LPP, Degeneracy in LPP, Sensitivity Analysis- I, Sensitivity Analysis- II, and Problems on Sensitivity Analysis. | CO2 |
| | B | Introduction to Duality Theory- I, Introduction to Duality Theory- II, Dual Simplex Method and Examples on Dual Simplex Method. | CO2 |
| | С | Integer Linear Programming, IPP: Branch & B-Bound Method and Mixed Integer Programming Problem. | CO2 |
| | Unit 3 | Introduction to transportation problem and Some Solving Methods | |



| А | Introduction to transportation problem-I, Transportation problem- II, Vogel Approximation method, optimal solution Generation for Transportation problem and Degeneracy in TP and problems. | CO3 |
|---------------------------|---|-----|
| В | Introduction to Nonlinear Programming, Graphical Solution of NLP, and Types of NLP. | CO3 |
| С | One-dimensional unconstrained optimization, Region Elimination Technique-1, Region Elimination Technique-2, and Region Elimination Technique-3. | CO3 |
| Unit 4 | NLP and Unconstrained optimization | |
| А | Multivariate Unconstrained Optimization-1, Multivariate Unconstrained Optimization-2. | CO4 |
| В | NLP with Equality Constrained-1, NLP with Equality Constrained-2, Constrained NLP-1, and Constrained NLP 2. | CO4 |
| С | Constrained Optimization, Constrained Optimization, and KKT (Karush-Kuhn-Tucker conditions) | CO4 |
| Unit 5 | Constrained optimization and Dynamic programming of LPP | |
| Α | Constrained Optimization, Constrained Optimization, and Feasible Direction. | CO5 |
| В | Penalty and barrier method, Penalty method, and Penalty and barrier method. | CO5 |
| С | Dynamic programming, Multi-Objective decision-making, and Multi-Attribute decision-making. | CO6 |
| Mode of examination | Theory | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | Kanti Swarup, P.K. Gupta, Man Mohan, Operations Research, Sultan Chand & Sons. S. Chandra, Jayadeva, Aparna Mehra, Numerical Optimization with Applications, Narosa. | |
| Other References | I Hamdy A. Taha, Operations Research, An Introduction, 9th Edition, Pearson. 2.M.S. Bazarra, H.D. Sheral, and C.M. Shetty, Nonlinear Programming Theory and Algorithms. | |

| РО | РО | PO | РО | PO | PO | РО | РО | РО | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA215.1 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 2 | 3 | | 1 | | | 1 |
| BDA215.2 | 1 | 2 | 3 | 2 | 1 | 1 | 1 | 2 | 3 | | 1 | | | 1 |
| BDA215.3 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 2 | 3 | | 1 | | | 1 |
| BDA215.4 | 1 | 2 | 3 | 3 | 1 | 1 | 1 | 2 | 3 | | 1 | | | 1 |
| BDA215.5 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 2 | 3 | | 1 | | | 1 |
| BDA215.6 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 2 | 3 | | 1 | | | 1 |
| Average | 1.0 | 2.0 | 2.3 | 2.1 | 1.0 | 1.0 | 1.0 | 2.0 | 3.0 | | 1.0 | | | 1.0 |



| Scho | ool: SSBSR | Batch: 2024-28 | |
|------|------------------------------|--|---|
| | gramme: B.Sc. | Academic Year: 2025-26 | |
| (Hor | | | |
| | nch: Data Science | Semester: III | |
| | nalytics Course Code | BDA216 | |
| 1 | | | |
| 2 | Course Title | Statistical Inference | |
| 3 | Credits | 4 | |
| 4 | Contact Hours (L-T-P) | 4-0-0 | |
| | Course Status | CC | |
| 5 | Course Objective | To introduce concepts of statistical analysis of descriptive statistics, le analytical tools, analyze and communicate quantitative data verbally, symbolically, and numerically. To make students familiar with the concept of Probability and S hypothesis. | graphically, |
| 6 | Course Outcomes Course | CO1: Describe the process of statistical analysis of descriptive principle of least square, lines of regression, simple linear regression multiple linear regression, coefficient of multiple determination. (K2, CO2: Describe the process of fitting polynomials and exponential cur CO3: Explain the criteria for obtaining a good estimator. (K2, K3) CO4: Calculate and interpret the point estimation, confidence construction of confidence intervals using a pivotal, shortest ex confidence interval. (K2, K3) CO5: Understand the null hypothesis, alternative hypothesis, type I error, level of significance, p-value, and power of the test, and deve to use a one-sample t-test, two-sample t-test, and paired-sample t- tests based on normal distribution one-sample and two-sample proble CO6: Develop the skills to interpret the results of statistical analysis Z-test, F-test, and chi-square test for goodness of fit. One-way a analysis of variance (ANOVA) techniques. (K2, K5) This is an advanced course in statistics. Students are introduced to the | , and evaluate K5) ves. (K2) interval, and pected length error, type II lop the ability test. Variance ms. (K2, K5) by using the nd Two-way |
| / | Description | involved in using sample data to make inferences about populations. the study of measures of central tendency and dispersion, finite statistical inferences from large and small samples, linear regr correlation and hypothesis. | Included are probability, ression, and |
| 8 | Outline syllabus | | CO Manaina |
| | Unit 1 | | Mapping CO1 |
| | A | Statistical analysis of descriptive statistics, the principle of least square, lines of regression, simple linear regression | COI |
| | В | Coefficient of determination. Multiple linear regression, coefficient of multiple determination. | CO2 |
| | С | Fitting of polynomials and exponential curves. | |
| | Unit 2 | | CO3 |
| | A | Criteria for obtaining a good estimator: unbiasedness, consistency, efficiency, and sufficiency. | CO3 |
| | В | Minimal sufficient statistic. | CO3 |
| | С | Uniformly minimum variance unbiased estimator, complete statistic. | |
| | Unit 3 | | CO4 |
| | А | Method of point estimation: Method of moments, maximum likelihood estimator, and its properties mean square error (MSE). | CO4 |



| В | Method of minimum chi-square, method of moments, Least square and their properties. | CO4 |
|---------------------------|--|-----|
| С | Interval estimation: Confidence interval, construction of confidence intervals | |
| Unit 4 | | CO5 |
| Α | Null hypothesis, alternative hypothesis, type I error, type II error, level of significance, p-value, and power of the test. | CO5 |
| В | Tests for mean based on normal distribution- one-sample t-test, two-sample t-test, paired-sample t-test. | CO5 |
| С | Tests for variance based on normal distribution- one-sample and two-sample problem | |
| Unit 5 | | CO6 |
| А | The large sample size test: Z-test, F-test, and Chi-square test for goodness of fit. | CO6 |
| В | One-way and Two-way analysis of variance (ANOVA) techniques. | CO6 |
| С | Statistical analysis of descriptive statistics, the principle of least square, lines of regression, simple linear regression | CO1 |
| Mode of examination | Theory | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | 1. Gupta, S.C. and Kapoor, V.K., "Fundamental of Mathematical Statistics". | |
| Other References | 1.Grewal, B.S, "Higher Engineering Mathematics".2.Goon, A.M., Gupta, A.K. & Das Gupta. Fundamental of Statistics. | |

| РО | PO | PO | РО | PO | PO | PO | PO | PO | PO | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA216.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | 1 | 1 |
| BDA216.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 1 | 1 | 1 | 1 |
| BDA216.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 1 | 1 | 1 | 1 |
| BDA216.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | 1 | 1 |
| BDA216.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | 1 | 1 |
| BDA216.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 1 | 1 | 1 | 1 |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | 1.0 | 1.0 | 1.0 |



| Scho | ool: SSBSR | Batch: 2024-28 | |
|------|---|---|---|
| | gramme: B.Sc. | Academic Year: 2025-26 | |
| (Hoi | ns.) nch: Data Science | Somestan III | |
| | nalytics | Semester: III | |
| 1 | Course Code | RBL001 | |
| 2 | Course Title | Research Based Learning-1 | |
| 3 | Credits | 0 | |
| 4 | Contact Hours (L-T-P) | 0-0-2 | |
| | Course Status | Compulsory | |
| 5 | Course Objective | Deep knowledge of a specific area of specialization. Develop communication skills, especially in project v presentation. Develop some time management skills. | vriting and oral |
| 6 | Course Outcomes | CO1: Explain the concept of research within the subject, as regard question, collecting and analyzing background material, and propressions and conclusions. (K2, K4) CO2: Construct and develop a deeper interest in mathematicates research. (K5, K6) CO3: Select and recommend activities that support their profession (K6) CO4: Develop effective project organizational skills. (K5) CO5: Analyse the problem and summarize research findings. (K4) CO6: Use research findings to develop education theory and practication. | resenting research s and a taste for sional goals. (K4, ,K5) |
| 7 | Course Description | Maintain a core of mathematical and technical knowledge that is a changing technologies and provides a solid foundation for future l | |
| 8 | | | |
| | Unit 1 | Introduction | CO1 |
| | | | |
| | Unit 2 | Case study | CO1,CO2 |
| | Unit 3 | Conceptual | CO2,CO3 |
| | Unit 4 | Development | CO3 |
| | Unit 5 | Finalisation | CO3,CO4 |
| | Mode of examination Weightage Distribution | | |
| | Text book/s* | | |
| | Other References | | |



| РО | РО | PO | PO | PO | PO | PO | PO | PO | РО | РО | PO | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| RBL001.1 | 3 | 3 | 2 | 2 | 1 | 1 | | | | | 2 | 1 | | |
| RBL001.2 | 2 | 3 | 2 | 2 | 1 | 1 | | | | | 2 | 1 | | |
| RBL001.3 | 2 | 2 | 2 | 3 | 1 | 1 | | | | | 2 | 1 | | |
| RBL001.4 | 2 | 3 | 2 | 2 | 1 | 1 | | | | | 2 | 1 | | |
| RBL001.5 | 3 | 3 | 2 | 2 | 1 | 1 | | | | | 2 | 1 | | |
| RBL001.6 | 3 | 3 | 2 | 3 | 1 | 1 | | | | | 2 | 1 | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | 1.0 | 1.0 | | | | | 2.0 | 1.0 | | |



| Scho | ool: SSBSR | Batch: 2024-28 | |
|------|-------------------|---|----------------|
| | gramme: B.Sc. | Academic Year: 2025-26 | |
| (Hor | | | |
| | nch: Data Science | Semester: III | |
| | nalytics | DD 4 415 | |
| 1 | Course Code | BDA217 | |
| 2 | Course Title | Data Preparation and Data Cleaning | |
| 3 | Credits | 3 | |
| 4 | Contact Hours | 3-0-0 | |
| | (L-T-P) | | |
| | Course Status | CC | |
| 5 | Course | To make students familiar with the concepts of preparing your da | ta; Working |
| | Objective | with dates and times, Data Cleaning, Data Structure, and cleaning Tex | kt Data. |
| 6 | Course | CO1: Describe preparing data: Rearranging and removing variables, r | renaming |
| | Outcomes | variables, Variable classes, calculating new numeric variables, and ex | |
| | | to Dividing a continuous variable into categories, and working with fa | actor |
| | | variables. (K1, K3) | |
| | | CO2: Discuss how to work with dates and times, adding and removin | |
| | | observations and explain about removing duplicate observations, sele of the data, selecting a random sample from a dataset, and sorting a dataset. | |
| | | K3, K4) | alasel. (K2, |
| | | CO3: Explain the data cleaning and technical representation of data. (| K2 K3 K4) |
| | | CO4: Discuss the data structure. (K2, K6) | 112, 113, 111) |
| | | CO5: Describe Character Normalization, Encoding Conversion and U | nicode |
| | | Normalization, Character Conversion, and Transliteration. (K1, K2) | |
| | | CO6: Discuss and evaluate Generating Regular Expressions in R, Co | mmon String |
| | | Processing Tasks in R, Approximate Text Matching, String Me Metrics, and Approximate Text Matching in R. | etrics, String |
| 7 | Course | This course introduces preparing your data; Working with dates and | |
| | Description | Cleaning, Data Structure, and cleaning Text Data. | , |
| | | | |
| 8 | | | |
| | Unit 1 | | |
| | A | Preparing your data: Rearranging and removing variables, renaming | CO1 |
| | | variables, Variable classes, Calculating new numeric variables, Dividing a continuous variable into categories, Working with factor | COI |
| | D | variables, | CO1 |
| | B C | Manipulating character variables: Concatenating character strings, | |
| | | extracting a substring, Searching a character variable. | CO1 |
| | Unit 2 | | |
| | А | Working with dates and times, Adding and removing observations, | CO2 |
| | В | Removing duplicate observations, Selecting a subset of the data, | CO2 |
| | С | Selecting a random sample from a dataset, Sorting a dataset. | CO2 |
| | Unit 3 | | |
| | А | Data Cleaning: The Statistical Value Chain, Raw Data, Input Data, Valid Data Statistics, and Output | CO3 |
| | В | Valid Data, Statistics, and Output. Technical Representation of Data: Numeric Data, Integers, | |
| | D | Integers in R, Real Numbers, Double Precision Numbers, The | CO3 |
| | | Concept of Machine Precision, Consequences of Working with | 005 |
| | | Floating Point Numbers, Dealing with the Consequences, Numeric Data in R, Text Data, Terminology and Encodings, | |
| | С | Unicode, Textual Data in R: Objects of Class Character, Encoding | |
| | | | |
| | | in R, Reading, and Writing of Data with Non-Local Encoding, | CO3C |
| | | in R, Reading, and Writing of Data with Non-Local Encoding, Detecting Encoding, Collation, and Sorting, Times and Dates. Time and Date Notation, Time and Date Storage in R, Time and | CO3C |



| | Saving Times. | |
|--------------|--|-----|
| Unit 4 | | |
| A | Data Structure: Introduction, Tabular Data, data.frame, Databases, dplyr, Matrix Data, Time Series, | CO4 |
| В | Graph Data, Web Data, Web Scraping, Web API, Other Data, Tidying Tabular Data, | CO4 |
| С | Variable Per Column, Single Observation Stored in Multiple Tables. | CO4 |
| Unit 5 | | |
| А | Cleaning Text Data: Character Normalization, Encoding Conversion and Unicode Normalization, Character Conversion and Transliteration, | CO5 |
| В | Pattern Matching with Regular, Expressions, Basic Regular Expressions, Practical Regular Expressions, Generating Regular Expressions in R, | CO5 |
| С | Common String Processing Tasks in R, Approximate Text Matching, String Metrics, String Metrics, and Approximate Text Matching in R. | CO6 |
| Mode of | Theory | |
| examination | | |
| Weightage | | |
| Distribution | CA:25%; ESE:75% | |
| Text book/s* | 1. Bad Data Handbook: Cleaning Up the Data So You Can Get Back | |
| | to Work by Q. Ethan McCallum 2. Best Practices in Data Cleaning: A Complete Guide to Everything You Need to Do Before and After Collecting Your Data by Jason W Osborne | |
| Other | 1. Data Wrangling with Python by Jacqueline Kazil | |
| References | 2. Principles of Data Wrangling: Practical Techniques for Data | |
| | Preparation by Tye Rattenbury | |

| РО | РО | PO | РО | PO | РО | РО | РО | РО | РО | РО | РО | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|-----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA217.1 | | 2 | 1 | 2 | | 1 | | 3 | | | 2 | | 1 | 1 |
| BDA217.2 | | 2 | 1 | 2 | | 1 | | 3 | | | 2 | | 1 | 1 |
| BDA217.3 | | 2 | 1 | 2 | | 1 | | 3 | | | 2 | | 1 | 1 |
| BDA217.4 | | 2 | 1 | 2 | | 1 | | 3 | | | 2 | | 1 | 1 |
| BDA217.5 | | 2 | 1 | 2 | | 1 | | 3 | | | 2 | | 1 | 1 |
| BDA217.6 | | 2 | 1 | 2 | | 1 | | 3 | | | 2 | | 1 | 1 |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 2.0 | | 1.0 | 1.0 |



| Scho | ool: SSBSR | Batch: 2024-28 | |
|------|-------------------------------|---|---|
| | gramme: B.Sc. | Academic Year: 2025-26 | |
| (Hor | | | |
| | ۱ch: Data Science ک lytics | & Semester: III | |
| 1 | Course Code | VOM203 | |
| 2 | Course Title | Basic Excel Modelling | |
| 3 | Credits | 3 | |
| 4 | Contact Hours | | |
| - | (L-T-P) | 0-0-6 | |
| | Course Status | SEC | |
| 5 | Course | 1. To use advanced formula techniques and sophisticated lookups | |
| U | Objective | 2. To distinguish between different functions. | |
| | | 3. To understand the pitfalls and strengths of commonly used funct | tions and to |
| | | apply correct functions to their Excel models. | lions, and to |
| 6 | Course | CO1: Select functionalities like Goal Seek, Data Tables, and t | he Scenario |
| | Outcomes | Manager to make your models more robust and identify uses of ma CO2: Explain creating and maintaining accurate, flexible, responsitive friendly spreadsheets. CO3: Construct automated tasks using functions, and make sure the clean dynamically. CO4: Examine array capabilities and explores a range of function dynamic lookup ranges. CO5: Explain data through graphs and charts, create data mode interactivity. CO6: Create visualizations to analyze and present data. | ve, and user- ne data stays ons to create |
| 7 | Course | In offices all throughout the world, spreadsheet software continues | to be one of |
| | Description | the most frequently used programs. A significant tool will be ad employability profile after you learn to use this software with assur day, there are millions of job postings in India alone that mention h abilities. Digital skills contribute to higher income and better of chances. | ance. Every aving Excel employment |
| 8 | Outline syllabus | | CO Mapping |
| | Unit 1 | Data Modeling and Macros | mapping |
| | A | Modelling Functions: SUMPRODUCT | CO1 |
| | B | Data Tables, Goal Seek, Scenario Manager, Solver. | CO1 |
| | B C | Record a Macro, run a Macro, edit a Macro, working with | |
| | | Macros, Relative Reference Macros | CO1 |
| | Unit 2 | Spreadsheet Design and Documentation | |
| | A | Spreadsheet Design Principles | CO2 |
| | В | Calculations, Interface and Navigation | CO2 |
| | С | Tables and Structured Referencing, Using Functions to Sort Data, Introduction to Array Formulas, working with an Array Function (TRANSPOSE), Solving Problems with Array Formulas. | CO2 |
| | Unit 3 | Data Cleaning and Preparation | |
| | А | Replace blanks with repeating values | CO3 |
| | В | Fix Dates (DATE, MONTH, YEAR, DAY, TEXT) | CO3 |
| | С | Remove Unwanted Spaces (TRIM, CLEAN), Diagnostic Tools (ISNUMBER, LEN, CODE), Remove Unwanted Characters (SUBSTITUTE, CHAR, VALUE) | CO3 |
| | Unit 4 | Building Professional Dashboards using Financial Functions and Advanced Lookups | CO4 |
| | A | Working with Dates (EOMONTH, EDATE, WORKDAY.INTL), Financial Functions (FV, PV, PMT), Loan Schedule (PMT, EDATE), Net Present Value and Internal Rate of Return (NPV, IRR), Depreciation Functions (SLN, SYD, DDB). | CO4 |



| | INDIRECT, ADDRESS, Introduction to OFFSET, Solving Problems with OFFSET. | CO4 |
|---------------------|--|-----|
| | Dashboard Design, Prepare Data, Construct Dashboard, Creative Charting, Interactive Dashboard | CO5 |
| Unit 5 | Data Analysis | |
| А | Correlation, Histogram, Multiple Correlation | CO5 |
| В | Regression, ANOVA, Rank, and Percentile | CO6 |
| С | Sampling, t-test, z-test | CO6 |
| Mode of | Practical Based | |
| examination | | |
| Weightage | CA: 25%; CE: 25%; ETE: 50% | |
| Distribution | CA. 2570, CL. 2570, ETE. 5070 | |
| Text book/s* | Michael Alexander, Excel® Dashboards & Reports for Dummies, John Wiley & Sons, Inc, ISBN: 978-1-119-07676-6, 2016. | |
| Other References | 1. Michael Alexander and Dick Kusleika, Excel 2016 Formulas, John Wiley & Sons, Inc, ISBN: 978-1-119-06786-3, 2016. | |

| РО | РО | PO | PO | PO | PO | PO | PO | PO | РО | РО | PO | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| VOM203.1 | | 2 | 1 | 2 | | 1 | 2 | 3 | 2 | | 2 | | | 1 |
| VOM203.2 | | 2 | 1 | 2 | | 1 | 2 | 3 | 2 | | 2 | | | 1 |
| VOM203.3 | | 2 | 1 | 2 | | 1 | 2 | 3 | 2 | | 2 | | | 1 |
| VOM203.4 | | 2 | 1 | 2 | | 1 | 2 | 3 | 2 | | 2 | | | 1 |
| VOM203.5 | | 2 | 1 | 2 | | 1 | 2 | 3 | 2 | | 2 | | | 1 |
| VOM203.6 | | 2 | 1 | 2 | | 1 | 2 | 3 | 2 | | 2 | | | 1 |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | 2.0 | 3.0 | 2.0 | | 2.0 | | | 1.0 |



| Scho | ool: SSBSR | Batch: 2024-28 | | | | | | | | |
|------|--|--|-----------------|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2025-26 | | | | | | | | |
| (Hoi | | | | | | | | | | |
| | nch: Data Science | Semester: III | | | | | | | | |
| | nalytics | | | | | | | | | |
| 1 | Course Code | ARP207 | | | | | | | | |
| 2 | Course Title | Logical Skill Building and Soft Skills | | | | | | | | |
| 3 | Credits | 2 | | | | | | | | |
| 4 | Contact Hours (L-T-P) | 0-1-2 | | | | | | | | |
| | Course Status | AEC | | | | | | | | |
| 5 | Course Objective To enhance the holistic development of students and improve their employabilit skills. To provide a 360-degree exposure to learning elements of the Busines English readiness program, behavioral traits, achieve softer communication level and a positive self-branding along with augmenting numerical and altitudina abilities. To step up skill and upgrade students across varied industry needs t enhance employability skills. By the end of this semester, a student will hav entered the threshold of his/her 1 st phase of employability enhancement and skil building activity exercise. | | | | | | | | | |
| 6 | Course | After completion of this course, students will be able to: | | | | | | | | |
| 0 | Outcomes | CO1: Ascertain a competency level through Building Essential Lang Skills | | | | | | | | |
| 7 | CO2: Build positive emotional competence in self and learn GOA SMART Goals techniques CO3: Apply positive thinking, goal setting and success-focused Management, which would help them in their academic as well a career CO4: Acquire satisfactory competency in use of aptitude, logical reasoning CO5: Develop strategic thinking and diverse mathematical cor building number puzzles CO6: Demonstrate an ability to apply various quantitative apti making business decisions | | | | | | | | | |
| / | Course Description | This Level 1 blended training approach equips the students for Indust employment readiness and combines elements of soft skills and nume to achieve this purpose. | | | | | | | | |
| 8 | | | | | | | | | | |
| | Unit 1 | | | | | | | | | |
| | A | Know Yourself: Core Competence. A very unique and interactive approach through an engaging questionnaire to ascertain a student's current skill level to design, architect, and expose a student to the right syllabus and also to identify the correct TNI/TNA levels of the student. | CO1 | | | | | | | |
| | В | Techniques of Self-Awareness Self-Esteem & Effectiveness Building Positive Attitude Building Emotional Competence | CO1, CO2 | | | | | | | |
| | С | Positive Thinking & Attitude Building Goal Setting and SMART Goals – Milestone Mapping Enhancing L S R W G and P (Listening Speaking Reading Writing Grammar and Pronunciation) | CO1, CO2,CO3 | | | | | | | |
| | Unit 2 | Introduction to APTITUDE TRAINING- Reasoning- Logical/ Analytical | | | | | | | | |
| | А | Syllogism Letter Series Coding, Decoding, Ranking & Their Comparison Level-1 | CO4 | | | | | | | |



| В | Number Puzzles | CO5 |
|---------------------------|---|-----|
| С | Selection Based On Given Conditions | CO5 |
| Unit 3 | Quantitative Aptitude | |
| А | Number Systems Level 1 Vedic Maths Level-1 | CO6 |
| В | Percentage, | CO6 |
| С | Ratio & Proportion Mensuration - Area & Volume Algebra | |
| Unit 4 | Verbal Abilities – 1 | CO1 |
| А | Reading Comprehension | CO2 |
| В | Spotting the Errors | |
| Unit 5 | Time & Priority Management | CO3 |
| А | Steven Covey Time Management Matrix | CO3 |
| В | Creating Self Time Management Tracker | |
| Mode of examination | | |
| Weightage Distribution | Class Assignment/Free Speech Exercises / JAM – 60% Group Presentations/Mock Interviews/GD/ Reasoning, Quant & Aptitude – 40% | |
| Text book/s* | Wiley's Quantitative Aptitude-P Anand Quantum CAT – Arihant Publications Quicker Maths- M. Tyra Power of Positive Action (English, Paperback, Napoleon Hill) Streets of Attitude (English, Paperback, Cary Fagan, Elizabeth Wilson) The 6 Pillars of self-esteem and awareness – Nathaniel Brandon Goal Setting (English, Paperback, Wilson Dobson | |
| Other | | |
| References | | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | PO | РО | РО | PSO | PSO | PSO |
|----------|----|----|-----|-----|-----|-----|-----|-----|-----|----|----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| ARP207.1 | | | 2 | 2 | 1 | 3 | 1 | 3 | 1 | | | | | |
| ARP207.2 | | | 3 | 2 | 1 | 3 | 1 | 3 | 1 | | | | | |
| ARP207.3 | | | 2 | 2 | 1 | 3 | 1 | 3 | 1 | | | | | |
| ARP207.4 | | | 2 | 2 | 1 | 3 | 1 | 3 | 1 | | | | | |
| ARP207.5 | | | 2 | 2 | 1 | 3 | 1 | 3 | 1 | | | | | |
| ARP207.6 | | | 2 | 2 | 1 | 3 | 1 | 3 | 1 | | | | | |
| Average | | | 2.0 | 2.0 | 1.0 | 3.0 | 1.0 | 3.0 | 1.0 | | | | | |



| Scho | ol: SSBSR | Batch: 2024-28 | | | | | | | |
|------|--|---|--|--|--|--|--|--|--|
| Prog | gramme: B.Sc. | Academic Year: 2025-26 | | | | | | | |
| (Hor | | | | | | | | | |
| | ich: Data Science | Semester: III | | | | | | | |
| | nalytics | | | | | | | | |
| 1 | Course Code | BDA261 | | | | | | | |
| 2 | Course Title | Statistical Inference Lab | | | | | | | |
| 3 | Credits | 1 | | | | | | | |
| 4 | Contact Hours (L-T-P) | 0-0-2 | | | | | | | |
| | Course Status | CC | | | | | | | |
| 5 | Course | To introduce concepts of statistical analysis of descriptive statistics, lo | gics, and | | | | | | |
| | Objective | analytical tools, analyze and communicate quantitative data verbally, g | | | | | | | |
| | 3 | symbolically, and numerically. To make students familiar with the concept of Probability and S | | | | | | | |
| | | hypothesis. | | | | | | | |
| 6 | Course Outcomes | CO1: Describe the process of statistical analysis of descriptive s principle of least square, lines of regression, simple linear regression, multiple linear regression, coefficient of multiple determination. (K2, I CO2: Describe the process of fitting of polynomials and exponential cu | and evaluate (X5) | | | | | | |
| | CO3: Explain the criteria for obtaining a good estimator. (K2, K3) CO4: Calculate and interpret the point estimation, confidence construction of confidence intervals using a pivotal, shortest exp confidence interval. (K2, K3) | | | | | | | | |
| | | CO5: Understand the null hypothesis, alternative hypothesis, type I error, level of significance, p-value, and power of the test, and develot to use a one-sample t-test, two-sample t-test, and paired-sample t-test variance based on normal distribution – one-sample and two-sample p K5) CO6: Develop the skills to interpret the results of statistical analysis I Z-test, F-test, and Chi-square test for goodness of fit. One-way an analysis of variance (ANOVA) techniques. (K2, K5) | op the ability est. Tests for roblem. (K2, oy using the | | | | | | |
| 7 | Course Description | This is an advances course in statistics. Students are introduced to the involved in using sample data to make inferences about populations. It the study of measures of central tendency and dispersion, finite statistical inferences from large and small samples, linear regree correlation and hypothesis. | ncluded are probability, | | | | | | |
| 8 | Outline syllabus | | СО | | | | | | |
| | • | | Mapping | | | | | | |
| | Unit 1 | Lab. Experiment 1 | | | | | | | |
| | A, B, C | Problem-based on the principle of least square, Simple linear regression, Multiple linear regression | CO1 | | | | | | |
| | Unit 2 | Lab. Experiment 2 | | | | | | | |
| | A, B, C | Problem-based on obtaining a good estimator: Unbiasedness, Consistency, Efficiency, Sufficiency. | CO2 | | | | | | |
| | Unit 3 | Lab. Experiment 3 | | | | | | | |
| | A, B, C | Problem-based on Point and Interval Estimation. | CO3 | | | | | | |
| | Unit 4 | Lab. Experiment 4 | | | | | | | |
| | A, B, C | Problem-based on Hypothesis Testing. | CO4 | | | | | | |
| | Unit 5 | Lab. Experiment 5 | | | | | | | |
| | A, B, C | Problem-based on One-way and Two-way analysis of variance (ANOVA) techniques. | CO5, CO6 | | | | | | |



| Mode of | Practical+Viva | |
|--------------|--|--|
| examination | | |
| Weightage | CA:25%; CE:25%; ESE:50% | |
| Distribution | CA.2570, CE.2570, ESE.5070 | |
| Text book/s* | 1. Goon A.M., Gupta M.K. and Dasgupta B. (2008): Fundamentals of | |
| | Statistics, World Press. | |
| Other | 1. Grewal, B.S, "Higher Engineering Mathematics". | |
| References | | |

| РО | PO | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA261.1 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | 1 | 2 | |
| BDA261.2 | 1 | 2 | 3 | 2 | | 1 | 1 | 3 | 1 | | 1 | 1 | 2 | |
| BDA261.3 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | 1 | 2 | |
| BDA261.4 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | 1 | 2 | |
| BDA261.5 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | 1 | 2 | |
| BDA261.6 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 1 | 1 | 2 | |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 1.0 | 1.0 | 2.0 | |



| Sch | ool: SSBSR | Batch: 2024-28 | | | | | | | | | |
|-----|-------------------------------|--|---------------|--|--|--|--|--|--|--|--|
| (Ho | | Academic Year: 2025-26 | | | | | | | | | |
| | nch: Data Science & lytics | Semester: III | | | | | | | | | |
| 1 | Course Code | BDA262 | | | | | | | | | |
| 2 | Course Title | Data Preparation and Data Cleaning Lab | | | | | | | | | |
| 3 | Credits | 1 | | | | | | | | | |
| 4 | Contact Hours(L- T-P) | 0-0-2 | | | | | | | | | |
| | Course Status | CC | | | | | | | | | |
| 5 | Course Objective | To make students familiar with the concepts of preparing your data; Working with dates and times, Data Cleaning, Data Structure, and Cleaning Text Data | | | | | | | | | |
| 6 | Course Outcomes | Outcomes Variables, Variable classes, calculating new numeric variables, an how to Dividing a continuous variable into categories, Working variables. (K1, K3) CO2: Discuss how to work with dates and times, add and remove and explain about removing duplicate observations, selecting a s data, selecting a random sample from a dataset, and sorting a data K4) CO3: Explain the data cleaning and technical representation of da K4) CO4: Discuss the data structure. (K2, K6) CO5: Describe Character Normalization, Encoding Conversion a Normalization, Character Conversion, and Transliteration. (K1, K2) CO6: Discuss and evaluate Generating Regular Expressions in String Processing Tasks in R, Approximate Text Matching, St | | | | | | | | | |
| 7 | Course Description | String Metrics, and Approximate Text Matching in R. This course introduces preparing your data; Working with dates a Data Cleaning, Data Structure, and cleaning Text Data. | | | | | | | | | |
| 8 | Outline syllabus | Data Cleaning, Data Structure, and cleaning Text Data. | CO Mapping | | | | | | | | |
| | Unit 1 | Lab. Experiment 1 | | | | | | | | | |
| | A, B, C | Problem-based on data collection and source of error. | CO1, CO2 | | | | | | | | |
| | Unit 2 | Lab. Experiment 2 | | | | | | | | | |
| | A, B, C | Problem-based on screening, diagnosis, and treatment of data. | CO2, CO3 | | | | | | | | |
| | Unit 3 | Lab. Experiment 3 | | | | | | | | | |
| | A, B, C | Problem-based on missing value and record value. | CO3, CO4 | | | | | | | | |
| | Unit 4 | Lab. Experiment 4 | | | | | | | | | |
| | A, B, C | Problem-based on quality control procedure, and data Integration. | CO4, CO5 | | | | | | | | |
| | Unit 5 | Lab. Experiment 5 | | | | | | | | | |
| | A, B, C | Problem-based on tools and techniques for data cleaning. | CO5, CO6 | | | | | | | | |
| | Mode of examination | Practical + Viva | | | | | | | | | |
| | Weightage Distribution | CA:25%; CE:25%; ESE:50% | | | | | | | | | |
| | Text book/s* | Bad Data Handbook: Cleaning Up the Data So You Can Get Back to Work by Q. Ethan McCallum Best Practices in Data Cleaning: A Complete Guide to Everything You Need to Do Before and After Collecting Your Data by Jason W Osborne | | | | | | | | | |



| Other | 1. Data Wrangling with Python by Jacqueline Kazil | |
|------------|--|--|
| References | 2. Principles of Data Wrangling: Practical Techniques for Data | |
| | Preparation by Tye Rattenbury | |

| РО | РО | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA262.1 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| BDA262.2 | 1 | 2 | 3 | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| BDA262.3 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| BDA262.4 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| BDA262.5 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| BDA262.6 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | 1.0 | 1.0 | 1.0 | 3.0 | 1.0 | 1.0 | 3.0 | 1.0 | 2.0 | 3.0 |



| Sch | ool: SSBSR | Batch: 2024-28 | |
|-------------|---|--|--|
| Prog (Ho | gramme: B.Sc. ns.) | Academic Year: 2025-26 | |
| | nch: Data Science nalytics | Semester: IV | |
| 1 | Course Code | BDA218 | |
| 2 | Course Title | Data Ware Housing & Data Mining | |
| 3 | Credits | 3 | |
| 4 | Contact Hours (L-T-P) | 3-0-0 | |
| | Course Status | | |
| 5 | Course Objective | Familiarise students with basic concepts of data warehousing, busidata mining, association rule mining and classification, clustering, data mining. | • |
| 6 | Course Outcomes | CO1: Discuss the Data warehousing Components, Cleanup, and trans Tools - Metadata. (K3, K5) CO2: Explain methods of business analysis, reporting, and query tool applications. (K2, K3, K4) CO3: Describe the OLAP guideline multidimensional versus multi re OLAP, categories of tools, OLAP tools, and the internet. (K2, K4) CO4: Explain and illustrate data mining functionalities, the interesting patterns, integration of a data mining system with data warehouse issu preprocessing. (K2, K3) CO5: Explain the basic concepts of decision tree induction, bayesian rule-based classification, classification by backpropagation and apply | s and lational gness of ues, and data classification, |
| | | vector machines, associative classification, lazy learners, other classif methods, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 | ication , K4, K6) |
| 7 | Course Description | vector machines, associative classification, lazy learners, other classif methods, and prediction. (K2, K3, K4) | fication , K4, K6) ng, business |
| 7 8 | | vector machines, associative classification, lazy learners, other classification, lazy learners, other classification, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clu trends in data mining. | fication , K4, K6) ng, business |
| - | Description | vector machines, associative classification, lazy learners, other classification, lazy learners, other classification, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clu trends in data mining. Data Warehousing | fication , K4, K6) ng, business ustering, and CO |
| - | Description Outline syllabus | vector machines, associative classification, lazy learners, other classification, lazy learners, other classification, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clu trends in data mining. | fication , K4, K6) ng, business ustering, and CO |
| - | Description Outline syllabus Unit 1 | vector machines, associative classification, lazy learners, other classification, lazy learners, other classification, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clu trends in data mining. Data Warehousing | ication , K4, K6) ng, business ustering, and CO Mapping |
| - | Description Outline syllabus Unit 1 A | vector machines, associative classification, lazy learners, other classification, lazy learners, other classification, end the state of the state of | fication , K4, K6) ng, business ustering, and CO Mapping CO1 |
| - | Description Outline syllabus Unit 1 A B | vector machines, associative classification, lazy learners, other classifimethods, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clutrends in data mining. Data Warehousing Data warehousing Components –Building a Data warehouse. Mapping the Data Warehouse to a Multiprocessor Architecture – DBMS Schemas for Decision Support | fication , K4, K6) ng, business ustering, and CO Mapping CO1 CO1 |
| - | Description Description Outline syllabus Unit 1 A B C | vector machines, associative classification, lazy learners, other classifimethods, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clutrends in data mining. Data Warehousing Data warehousing Components –Building a Data warehouse. Mapping the Data Warehouse to a Multiprocessor Architecture – DBMS Schemas for Decision Support Data Extraction, Cleanup, and Transformation Tools - Metadata. Business Analysis Reporting and Query tools and Applications, Cognos Impromptu, Online Analytical Processing (OLAP). | ication , K4, K6) ng, business ustering, and CO Mapping CO1 CO1 |
| - | Description Description Outline syllabus Unit 1 A B C Unit 2 A B | vector machines, associative classification, lazy learners, other classifimethods, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clutrends in data mining. Data Warehousing Data warehousing Components –Building a Data warehouse. Mapping the Data Warehouse to a Multiprocessor Architecture – DBMS Schemas for Decision Support Data Extraction, Cleanup, and Transformation Tools - Metadata. Business Analysis Reporting and Query tools and Applications, Cognos Impromptu, Online Analytical Processing (OLAP). Multidimensional Data Model, OLAP Guideline Multidimensional | ication , K4, K6) ng, business ustering, and CO Mapping CO1 CO1 CO1 CO1 CO2, CO3 |
| - | Description Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 2 A C C A | vector machines, associative classification, lazy learners, other classifimethods, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clutrends in data mining. Data Warehousing Data warehousing Components –Building a Data warehouse. Mapping the Data Warehouse to a Multiprocessor Architecture – DBMS Schemas for Decision Support Data Extraction, Cleanup, and Transformation Tools - Metadata. Business Analysis Reporting and Query tools and Applications, Cognos Impromptu, Online Analytical Processing (OLAP). Multidimensional Data Model, OLAP Guideline Multidimensional versus Multirotational OLAP, Categories of Tools, OLAP Tools, and the Internet. | ication , K4, K6) ng, business ustering, and CO Mapping CO1 CO1 CO1 CO2, CO3 CO3 |
| - | Description Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 2 A Unit 3 | vector machines, associative classification, lazy learners, other classifimethods, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clutrends in data mining. Data Warehousing Data warehousing Components –Building a Data warehouse. Mapping the Data Warehouse to a Multiprocessor Architecture – DBMS Schemas for Decision Support Data Extraction, Cleanup, and Transformation Tools - Metadata. Business Analysis Reporting and Query tools and Applications, Cognos Impromptu, Online Analytical Processing (OLAP). Multidimensional Data Model, OLAP Guideline Multidimensional versus Multirotational OLAP, Categories of Tools, OLAP Tools, and the Internet. Data Mining | ication , K4, K6) ng, business ustering, and CO Mapping CO1 CO1 CO1 CO2, CO3 CO3 CO3 CO3 |
| - | Description Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 2 A C C A | vector machines, associative classification, lazy learners, other classifimethods, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clutrends in data mining. Data Warehousing Data warehousing Components –Building a Data warehouse. Mapping the Data Warehouse to a Multiprocessor Architecture – DBMS Schemas for Decision Support Data Extraction, Cleanup, and Transformation Tools - Metadata. Business Analysis Reporting and Query tools and Applications, Cognos Impromptu, Online Analytical Processing (OLAP). Multidimensional Data Model, OLAP Guideline Multidimensional versus Multirotational OLAP, Categories of Tools, OLAP Tools, and the Internet. Data Mining Introduction, Data, Types of Data, Data Mining Functionalities, | ication , K4, K6) ng, business ustering, and CO Mapping CO1 CO1 CO1 CO2, CO3 CO3 |
| - | Description Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 3 A B | vector machines, associative classification, lazy learners, other classifimethods, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clutrends in data mining. Data Warehousing Data warehousing Components –Building a Data warehouse. Mapping the Data Warehouse to a Multiprocessor Architecture – DBMS Schemas for Decision Support Data Extraction, Cleanup, and Transformation Tools - Metadata. Business Analysis Reporting and Query tools and Applications, Cognos Impromptu, Online Analytical Processing (OLAP). Multidimensional Data Model, OLAP Guideline Multidimensional versus Multirotational OLAP, Categories of Tools, OLAP Tools, and the Internet. Data Mining Introduction, Data, Types of Data, Data Mining Functionalities, Interestingness of Patterns, Classification of Data Mining Systems, Data Mining Task Primitives. | ication , K4, K6) ng, business ustering, and CO Mapping CO1 CO1 CO1 CO2, CO3 CO3 CO3 CO3 |
| - | Description Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 2 A B C Unit 3 A | vector machines, associative classification, lazy learners, other classifimethods, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clutrends in data mining. Data Warehousing Data warehousing Components –Building a Data warehouse. Mapping the Data Warehouse to a Multiprocessor Architecture – DBMS Schemas for Decision Support Data Extraction, Cleanup, and Transformation Tools - Metadata. Business Analysis Reporting and Query tools and Applications, Cognos Impromptu, Online Analytical Processing (OLAP). Multidimensional Data Model, OLAP Guideline Multidimensional versus Multirotational OLAP, Categories of Tools, OLAP Tools, and the Internet. Data Mining Introduction, Data, Types of Data, Data Mining Functionalities, Interestingness of Patterns, Classification of Data Mining Systems, Data Mining Task Primitives, Integration of a Data Mining System with Data Warehouse Issues, Data Preprocessing | ication , K4, K6) ng, business ustering, and CO1 CO1 CO1 CO1 CO2, CO3 CO3 CO3 CO3 |
| - | Description Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 3 A B | vector machines, associative classification, lazy learners, other classifimethods, and prediction. (K2, K3, K4) CO6: Explain and evaluate clustering and trends in data mining. (K2 This course introduces the basic concepts of data warehousin analysis, data mining, association rule mining and classification, clutrends in data mining. Data Warehousing Data warehousing Components –Building a Data warehouse. Mapping the Data Warehouse to a Multiprocessor Architecture – DBMS Schemas for Decision Support Data Extraction, Cleanup, and Transformation Tools - Metadata. Business Analysis Reporting and Query tools and Applications, Cognos Impromptu, Online Analytical Processing (OLAP). Multidimensional Data Model, OLAP Guideline Multidimensional versus Multirotational OLAP, Categories of Tools, OLAP Tools, and the Internet. Data Mining Introduction, Data, Types of Data, Data Mining Functionalities, Interestingness of Patterns, Classification of Data Mining Systems, Data Mining Task Primitives, Integration of a Data Mining System with Data Warehouse | ication , K4, K6) ng, business ustering, and CO1 CO1 CO1 CO1 CO2, CO3 CO3 CO3 CO3 |



| | Methods, Mining various Kinds of Association Rules, Correlation Analysis, | |
|---------------------------|--|-----|
| В | Constraint-Based Association Mining Classification and Prediction, Basic Concepts, Decision Tree Induction, Bayesian Classification, Rule Based Classification, Classification by Backpropagation, | CO5 |
| С | Support Vector Machines, Associative Classification, Lazy Learners, Other Classification Methods, and Prediction. | |
| Unit 5 | Clustering and Trends in Data Mining | CO6 |
| А | Cluster Analysis, Types of Data, Categorization of Major Clustering Methods, K-means, Partitioning Methods, Hierarchical Methods, | CO6 |
| В | Density-Based Methods, Grid-Based Methods, Model-Based Clustering Methods, Clustering High Dimensional Data, Constraint, Based Cluster Analysis, and Outlier Analysis. | CO6 |
| С | Data Mining Applications. Apply data mining techniques and methods to large data sets, use data mining tools, and Compare and contrast the various classifiers. | |
| Mode of | Theory | |
| examination | | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | Alex Berson and Stephen J. Smith, "Data Warehousing, Data Mining and OLAP", Tata McGraw Hill Jiawei Han and Micheline Kamber, "Data Mining Concepts and Techniques", Third Edition, Elsevier. | |
| Other References | Pang-Ning Tan, Michael Steinbach, and Vipin Kumar, "Introduction to Data Mining", Person Education. K.P. Soman, Shyam Diwakar and V. Aja, "Insight into Data | |
| | Mining Theory and Practice", Eastern Economy Edition, Prentice Hall. | |

| РО | РО | PO | PO | PO | PO | PO | PO | PO | РО | РО | PO | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA218.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | | |
| BDA218.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 1 | 1 | | |
| BDA218.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 1 | 1 | | |
| BDA218.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | | |
| BDA218.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | | |
| BDA218.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 1 | 1 | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | 1.0 | | |



| Scho | ool: SSBSR | Batch: 2024-28 | | | | | | | | |
|-------------|-------------------------------|---|-------------------------------|--|--|--|--|--|--|--|
| Prog (Ho | gramme: B.Sc. | Academic Year: 2025-26 | | | | | | | | |
| Brai | nch: Data Science nalytics | Semester: IV | | | | | | | | |
| 1 | Course Code | BDA202 | | | | | | | | |
| 2 | Course Title | Database Management Systems | | | | | | | | |
| 3 | Credits | 4 | | | | | | | | |
| 4 | Contact Hours (L-T-P) | 4-0-0 | | | | | | | | |
| | Course Status | СС | | | | | | | | |
| 5 | Course Objective | To make students familiar with the basic concepts of Databases and and Data Models, Database Design, ER-Diagram and Unified Language, Relational Algebra and Calculus, Constraints, Views Transaction management, and Concurrency control. | d Modeling and SQL, | | | | | | | |
| 6 | Course Outcomes | CO1: Discuss the basics of Databases and Transactions and Data Mor K3) CO2: Discuss about Database Design, ER-Diagram, and Unific Language. (K1, K3) CO3: Explain relational algebra and calculus, describe Domain relation calculus vs algebra, and computational capabilities. (K3, K4) CO4: Explain and illustrate Constraints, Views, and SQL. (K3, K6) CO5: Evaluate different types of transaction management. (K4, K5) CO6: Explain concurrency control, time stamping methods, optimis and database recovery management. (K2, K4, K5) | ed Modeling onal Calculus, | | | | | | | |
| 7 | Course Description | This course introduces the basic concepts of Databases and Transactic Models, Database Design, ER-Diagram and Unified Modeling Relational Algebra and Calculus, Constraints, Views and SQL, management, and Concurrency control. | Language, | | | | | | | |
| 8 | Outline syllabus | | CO Mapping | | | | | | | |
| | Unit 1 | Introduction to Databases and Transactions and Data Models | | | | | | | | |
| | А | What is a database system, purpose of the database system, what view of data, relational databases, database architecture. | CO1 | | | | | | | |
| | В | Transaction management, The importance of data models, Basic building blocks, | CO1 | | | | | | | |
| | С | Business rules, The evolution of data models, Degrees of data abstraction. | CO1 | | | | | | | |
| | Unit 2 | Database Design, ER-Diagram, and Unified Modeling Language | | | | | | | | |
| | А | Database design and ER Model: overview, ER-Model, Constraints, ER-Diagrams, ERD Issues, weak entity sets, Codd's rules, Relational Schemas, | CO2 | | | | | | | |
| | В | Introduction to UML Relational database model: Logical view of data, keys, integrity rules. | CO2 | | | | | | | |
| | С | Relational Database design: features of good relational database design, atomic domain, and Normalization (1NF, 2NF, 3NF, BCNF). | CO2 | | | | | | | |
| | Unit 3 | Relational Algebra and Calculus | | | | | | | | |
| | А | Relational algebra: introduction, Selection, and projection, set operations, renaming, Joins, Division, syntax, semantics. | CO3 | | | | | | | |
| | В | Operators, grouping and ungrouping, relational comparison. | CO3 | | | | | | | |
| | С | Calculus: Tuple relational calculus, Domain relational Calculus, calculus vs algebra, computational capabilities. | CO3 | | | | | | | |
| | Unit 4 | Constraints, Views, and SQL | | | | | | | | |



| | A | What are constraints, types of constraints, and Integrity constraints? | CO4 |
|---|---------------------------|--|----------|
| | В | Views: Introduction to views, data independence, security, updates on views, and comparison between tables. | CO4 |
| | С | Views SQL: data definition, aggregate function, Null Values, nested subqueries, Joined relations. Triggers. | CO4 |
| | Unit 5 | Transaction management and Concurrency control | |
| | A | Transaction management: ACID properties, serializability, and concurrency control, | CO5, CO6 |
| | В | Lock-based concurrency control (2PL, Deadlocks), Time stamping methods. | CO5, CO6 |
| | С | Optimistic methods, database recovery management. | CO5, CO6 |
| - | Mode of | Theory | |
| | examination | | |
| | Weightage Distribution | CA:25%; ESE:75% | |
| , | Text book/s* | 1."Database System Concepts", 6th Edition by Abraham Silberschatz, Henry F. Korth, S. Sudarshan, McGraw-Hill | |
| , | Other | 1 "Principles of Database and Knowledge – Base Systems", Vol 1 | |
| | References | by J. D. Ullman, Computer science Press. | |
| | | 2 "Fundamentals of Database Systems", 5th Edition by R. Elmasri | |
| | | and S. Navathe, Pearson Education | |

| PO | РО | PO | РО | PO | PO | PO | PO | PO | PO | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA202.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | | |
| BDA202.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 1 | 1 | | |
| BDA202.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 1 | 1 | | |
| BDA202.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | | |
| BDA202.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | | |
| BDA202.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 1 | 1 | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | 1.0 | | |



| | ool: SSBSR | Batch: 2024-28 | |
|------|-------------------------------|--|---|
| (Hoi | | Academic Year: 2025-26 | |
| | nch: Data Science nalytics | | |
| 1 | Course Code | BDA214 | |
| 2 | Course Title | Sampling Theory | |
| 3 | Credits | 4 | |
| 4 | Contact Hours (L-T-P) | 4-0-0 | |
| | Course Status | DSE | |
| 5 | Course Objective | To make students familiar with the concept of sample and popul enumeration versus sampling. The concept of Systematic Samplin the population mean and total, variances of these estimates along we the present official statistical system in India, methods of collect statistics, their reliability, and limitations have been introduced. | ng, estimates of vith the brief of |
| 6 | Course Outcomes | CO1: Explain and illustrate the concepts of sample and population. CO2: Describe the properties of complete enumeration versus sat random sampling with and without replacement. (K1, K2, K3) CO3: Describe estimates of the population mean, explain its estimates of these variances, and sample size determination. (K2, K2 CO4: Describe stratified random sampling, estimates of the popul total and explain its application, and illustrate systematic sampling. CO5: Describe the ratio and regression methods of estimation variances in terms of the correlation coefficient between X and Y for method and their comparison with SRS. (K2, K3, K6) CO6: Describe and analyze the basic concepts present official statis India, and methods of collection of official statistics. (K1, K2, K4) | mpling; explain application and 3, K4) ation mean and (K2, K3, K4) n and evaluate or the regression |
| 7 | Course Description | This course initiates the advanced concept of sample and populat enumeration versus sampling. The concept of Systematic Sampling the population mean and total, variances of these estimates along we the present official statistical system in India, methods of collect statistics, their reliability, and limitations have been introduced. | ion, complete g, estimates of ith the brief of |
| 8 | | | |
| | Unit 1 | | |
| | А | Concept of sample and population, complete enumeration versus sampling | CO1 |
| | В | Sampling and non-sampling errors, requirements of a good sample, | CO1 |
| | С | Simple random sampling with and without replacement. | CO2 |
| | Unit 2 | | |
| | А | Estimates of the population mean, total, and proportion, | CO3 |
| | В | Variances of these estimates | CO3 |
| | С | Estimates of theses variances and sample size determination. | CO3 |
| | Unit 3 | | |
| | А | Stratified random sampling, estimates of the population mean, and total variances of these estimates. | CO4 |
| | В | Proportional and optimum allocations and their comparison with SRS. | CO4 |
| | С | Systematic Sampling, estimates of the population mean and total, variances of these estimates. | CO4 |
| | Unit 4 | | |
| | Α | Ratio and regression methods of estimation, estimates of the population mean and total (for SRS of large size), | CO5 |



| В | Variances of these estimates and estimates of theses variances, | CO5 |
|--------------|--|-----|
| | | |
| С | Variances in terms of the correlation coefficient between X and Y for regression method and their comparison with SRS. | CO5 |
| Unit 5 | | |
| А | Present official statistical system in India, Methods of collection of official statistics, their reliability and limitations. | CO6 |
| В | Principal publications containing data on the topics such as population, industry, and finance. | CO6 |
| С | Various official agencies are responsible for data collection and their main functions. | CO6 |
| Mode of | Theory | |
| examination | | |
| Weightage | | |
| Distribution | CA:25%; ESE:75% | |
| Text book/s* | 1. Murthy M.N. (1977): Sampling Theory & Statistical | |
| | Methods, Statistical Pub. Society, Calcutta | |
| | 2. Cochran W.G (1984): Sampling Techniques (3rd Ed.), Wiley | |
| | Eastern. | |
| Other | 1. Mukhopadhyay P. (1998): Theory and Methods of Survey | |
| References | Sampling, Prentice Hall | |
| | 2. Guide to Current Indian Official Statistics, Central Statistical | |
| | Organization, GOI, New Delhi. | |

| РО | РО | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA214.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | 1 | |
| BDA214.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 1 | 1 | 1 | |
| BDA214.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 1 | 1 | 1 | |
| BDA214.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | 1 | |
| BDA214.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 1 | 1 | 1 | |
| BDA214.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 1 | 1 | 1 | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 1.0 | 1.0 | 1.0 | |



| Sch | ool: SSBSR | Batch: 2024-28 | | | | | | | | | | | |
|-----|--|---|------------------|--|--|--|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2025-26 | | | | | | | | | | | |
| | ns.) nch: Data Science analytics | Semester: IV | | | | | | | | | | | |
| 1 1 | Course Code | RBL002 | | | | | | | | | | | |
| 2 | Course Title | Research Based Learning-2 | | | | | | | | | | | |
| 3 | Credits | 0 | | | | | | | | | | | |
| 4 | Contact Hours | 0-0-2 | | | | | | | | | | | |
| • | (L-T-P) | | | | | | | | | | | | |
| | Course Status | Project (Audit-Qualifying) | | | | | | | | | | | |
| 5 | Course | 1. Deep knowledge of a specific area of specialization. | | | | | | | | | | | |
| 5 | Objective | 2. Develop communication skills, especially in project with | riting and oral | | | | | | | | | | |
| | objective | presentation. Develop some time management skills. | 0 | | | | | | | | | | |
| 6 | Course | CO1: Explain the concept of research within the subject, as regard | ls approaching a | | | | | | | | | | |
| | Outcomes | question, collecting and analyzing background material, and pre | | | | | | | | | | | |
| | | questions and conclusions. (K2, K4) | | | | | | | | | | | |
| | | CO2: Construct and develop a deeper interest in mathematics | and a taste for | | | | | | | | | | |
| | | research. (K5, K6) | 1 1 (17.4 | | | | | | | | | | |
| | | CO3: Select and recommend activities that support their professi | onal goals. (K4, | | | | | | | | | | |
| | | K6) CO4: Develop offective project experimetional skills (K5) | | | | | | | | | | | |
| | | CO4: Develop effective project organizational skills. (K5) CO5: Analyse the problem and summarize research findings. (K4,k | (5) | | | | | | | | | | |
| | | CO6: Use research findings to develop education theory and practic | | | | | | | | | | | |
| 7 | Course | Maintain a core of mathematical and technical knowledge that is ad | | | | | | | | | | | |
| | Description | changing technologies and provides a solid foundation for future le | | | | | | | | | | | |
| 8 | | | | | | | | | | | | | |
| | Unit 1 | Introduction | CO1 | | | | | | | | | | |
| | Unit 2 | Case study | CO1,CO2 | | | | | | | | | | |
| | | | | | | | | | | | | | |
| | Unit 3 | Conceptual | CO2,CO3 | | | | | | | | | | |
| | | | | | | | | | | | | | |
| | Unit 4 | Development | CO3 | | | | | | | | | | |
| | Unit 5 | Finalisation | CO3,CO4 | | | | | | | | | | |
| | Mode of | | | | | | | | | | | | |
| | examination | | | | | | | | | | | | |
| | Weightage | | | | | | | | | | | | |
| | Distribution | | | | | | | | | | | | |
| | Text book/s* | | | | | | | | | | | | |
| | Other | | | | | | | | | | | | |
| | References | | | | | | | | | | | | |



| PO | РО | PO | PO | РО | РО | РО | РО | РО | PO | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| RBL002.1 | 3 | 3 | 2 | 2 | 1 | 1 | | | | | 1 | 1 | | |
| RBL002.2 | 2 | 3 | 3 | 2 | 1 | 1 | | | | | 1 | 1 | | |
| RBL002.3 | 2 | 2 | 2 | 3 | 1 | 1 | | | | | 1 | 1 | | |
| RBL002.4 | 2 | 3 | 2 | 2 | 1 | 1 | | | | | 1 | 1 | | |
| RBL002.5 | 3 | 3 | 2 | 2 | 1 | 1 | | | | | 1 | 1 | | |
| RBL002.6 | 3 | 3 | 2 | 3 | 1 | 1 | | | | | 1 | 1 | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | 1.0 | 1.0 | | | | | 1.0 | 1.0 | | |



| Sch | ool: SSBSR | Batch: 2024-28 | | | |
|-----|-----------------------------|--|---------------|--|--|
| | ogram: B.Sc. (Ho | | | | |
| | anch: Data Scien alytics | nce & Semester: IV | | | |
| 1 | Course Cod | le ARP306 | | | |
| 2 | Course Titl | e Campus to Corporate | | | |
| 3 | Credits | 2 | | | |
| 4 | Contact Hou (L-T-P) | 0-1-2 | | | |
| | Course State | | | | |
| 5 | Course Objec | tive To enhance holistic development of students and improve their employability skills. Provide a 360-degree exposure to learning elements of Business English readiness program, behavioural traits, achieve softer communication levels and a positive self- branding along with augmenting numerical and altitudinal abilities. To up skill and upgrade students across varied industry needs to enhance employability skills. By the end of this semester, a will have entered the threshold of his/her 4th phase of employability enhancement and skill building activity exercise. | | | |
| 6 | Course Outco | After completion of this course, students will be able to: CO1: Develop a creative resume, cover letters, interpret job descriptions and interpret KRA and KPI statements and art of conflict management. CO2: Build negotiation skills to get maximum benefits from deals in practical life scenarios. CO3: Develop skills of personal branding to create a brand image and self-branding | | | |
| 7 | Course Descrip | business decisions. This penultimate stage introduces the student to the basics of Human Resources. Allows the student to understand and interpret KRA KPI and understand Job descriptions. A student also understands how to manage conflicts, brand himself/herself understand relations and empathise others with level-4 of quant aptitude and logical reasoning | | | |
| 8 | | Outline syllabus – ARP 306 | | | |
| | Unit 1 | Ace the Interview | CO MAPPING | | |
| | Δ | HR Sensitization (Role Clarity KRA KPI Understanding JD) Conflict Management | CO1 | | |
| Ĺ | | Negotiation Skills Personal Branding | CO3, CO4 | | |
| | C . | Uploading & Curating Resumes in Job Portals, getting Your Resumes Noticed Writing Cover Letters Relationship Management | CO1, CO3 | | |
| | Unit 2 | Introduction to APTITUDE TRAINING- Reasoning- Logical/ | | | |



| | Analytical | |
|-------------------------|---|-----|
| А | Sitting Arrangement & Venn Diagrams Puzzles Distribution Selection | CO4 |
| В | Direction Sense Statement & Conclusion Strong & Weak Arguments | CO4 |
| С | Analogies, Odd One out Cause & Effect | CO5 |
| Unit 3 | Quantitative Aptitude | |
| А | Average, Ratio & Proportions, Mixtures & Allegation | CO6 |
| В | Geometry-Lines, Angles & Triangles | CO6 |
| С | Problem of Ages Data Sufficiency - L2 | CO6 |
| Unit 4 | Verbal Abilities-4 | |
| А | Antonyms and Synonyms | CO1 |
| В | Idioms and Phrases | CO2 |
| Unit 5 | Problem Solving and Case Studies | |
| А | Real time Case Study Solving Exercises | CO4 |
| В | Intra student Mock Situation Handling Exercises | CO4 |
| Evaluation Weightage | (CA)Class Assignment/Free Speech Exercises / JAM – 60% (ETE) Group Presentations/Mock Interviews(MIP's)/GD/ Reasoning, Quant & Aptitude– 40% | |
| Text book/s* | Wiley's Quantitative Aptitude-P Anand Quantum CAT – Arihant Publications Quicker Maths- M. Tyra Power of Positive Action (English, Paperback, Napoleon Hill) Streets of Attitude (English, Paperback, Cary Fagan, Elizabeth Wilson) The 6 Pillars of self-esteem and awareness – Nathaniel Brandon Goal Setting (English, Paperback, Wilson Dobson | |

| РО | РО | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|----|----|-----|-----|----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| ARP306.1 | | | 2 | 2 | | 3 | 1 | 3 | 1 | | 2 | | | |
| ARP306.2 | | | 3 | 2 | | 3 | 1 | 3 | 1 | | 2 | | | |
| ARP306.3 | | | 2 | 2 | | 3 | 1 | 3 | 1 | | 2 | | | |
| ARP306.4 | | | 2 | 2 | | 3 | 1 | 3 | 1 | | 2 | | | |
| ARP306.5 | | | 2 | 2 | | 3 | 1 | 3 | 1 | | 2 | | | |
| ARP306.6 | | | 2 | 2 | | 3 | 1 | 3 | 1 | | 2 | | | |
| Average | | | 2.0 | 2.0 | | 3.0 | 1.0 | 3.0 | 1.0 | | 2.0 | | | |



| | ool: SSBSR | Batch: 2024-28 | |
|-----|---|--|---|
| (Ho | | Academic Year: 2025-26 | |
| | nch: Data | Semester: IV | |
| | ence & Analytics | | |
| 1 | Course Code | BDA270 | |
| 2 | Course Title | Data Ware Housing & Data Mining Lab | |
| 3 | Credits | 1 | |
| 4 | Contact Hours (L-T-P) | 0-0-2 | |
| | Course Status | CC | |
| 5 | Course Objectiv e | To introduce students to basic applications, concepts, and mining. To develop skills for using recent data mining software (eg. R problems in a variety of disciplines. | - |
| 6 | Course Outcome s | 3. To gain experience doing independent study and research CO1: Learn how to build a data warehouse and query it (using ope Pentaho Data Integration Tool, Pentaho Business Analytics). (K2, CO2: Learn to perform data mining tasks using a data mining tool | (K5) |
| | | source WEKA). (K2) CO3: Understand the data sets and data preprocessing. (K2, K3) CO4: Demonstrate the working of algorithms for data mining task rule mining, classification, clustering and regression. (K2, K3) CO5: Exercise the data mining techniques with varied input value parameters. (K2, K5) | |
| | | | ets. (K2, K5) |
| 7 | Course Descriptio n | CO6: To obtain Practical Experience Working with all real data see To introduce students to basic applications, concepts, and technique To develop skills for using recent data mining software to solve in a variety of disciplines. To gain experience doing independent s | ues of data mining. practical problem |
| 7 8 | Descriptio | CO6: To obtain Practical Experience Working with all real data set To introduce students to basic applications, concepts, and technique To develop skills for using recent data mining software to solve in a variety of disciplines. To gain experience doing independent set | ues of data mining. practical problem |
| | Descriptio n | CO6: To obtain Practical Experience Working with all real data set To introduce students to basic applications, concepts, and technique To develop skills for using recent data mining software to solve in a variety of disciplines. To gain experience doing independent set | ues of data mining. practical problem study and research |
| | Descriptio n Outline syllabu Unit 1 A, B, C | CO6: To obtain Practical Experience Working with all real data set To introduce students to basic applications, concepts, and technique To develop skills for using recent data mining software to solve in a variety of disciplines. To gain experience doing independent set | ues of data mining. practical problem study and research |
| | Descriptio n Outline syllabu Unit 1 | CO6: To obtain Practical Experience Working with all real data set To introduce students to basic applications, concepts, and technique To develop skills for using recent data mining software to solve in a variety of disciplines. To gain experience doing independent set Is Lab. Experiment 1 Installation of WEKA Tool | ues of data mining practical problem study and research CO Mapping |
| | Descriptio n Outline syllabu Unit 1 A, B, C Unit 2 A, B, C | CO6: To obtain Practical Experience Working with all real data set To introduce students to basic applications, concepts, and technique To develop skills for using recent data mining software to solve in a variety of disciplines. To gain experience doing independent set IS Lab. Experiment 1 Installation of WEKA Tool Creating new Arff File Lab. Experiment 2 Pre-Processes Techniques on Data Set Pre-process a given dataset based on Handling Missing Values | ues of data mining practical problem study and research CO Mapping |
| | Descriptio n Outline syllabu Unit 1 A, B, C Unit 2 | CO6: To obtain Practical Experience Working with all real data setTo introduce students to basic applications, concepts, and techniqueTo develop skills for using recent data mining software to solvein a variety of disciplines. To gain experience doing independent setISLab. Experiment 1Installation of WEKA ToolCreating new Arff FileLab. Experiment 2Pre-Processes Techniques on Data SetPre-process a given dataset based on Handling Missing ValuesLab. Experiment 3 | ues of data mining practical problem study and research CO Mapping CO1 |
| | Descriptio n Outline syllabu Unit 1 A, B, C Unit 2 A, B, C Unit 3 A, B, C | CO6: To obtain Practical Experience Working with all real data setTo introduce students to basic applications, concepts, and techniqueTo develop skills for using recent data mining software to solvein a variety of disciplines. To gain experience doing independent setISLab. Experiment 1Installation of WEKA ToolCreating new Arff FileLab. Experiment 2Pre-Processes Techniques on Data SetPre-process a given dataset based on Handling Missing ValuesLab. Experiment 3Generate Association Rules using the Apriori AlgorithmGenerating association rules using fp growth algorithm | ues of data mining practical problem study and research CO Mapping CO1 |
| | Descriptio n Outline syllabu Unit 1 A, B, C Unit 2 A, B, C Unit 3 | CO6: To obtain Practical Experience Working with all real data setTo introduce students to basic applications, concepts, and techniqueTo develop skills for using recent data mining software to solvein a variety of disciplines. To gain experience doing independent setISLab. Experiment 1Installation of WEKA ToolCreating new Arff FileLab. Experiment 2Pre-Processes Techniques on Data SetPre-process a given dataset based on Handling Missing ValuesLab. Experiment 3Generate Association Rules using the Apriori AlgorithmGenerating association rules using fp growth algorithmLab. Experiment 4 | ues of data mining practical problem study and research CO Mapping CO1 CO2 |
| | Descriptio n Outline syllabu Unit 1 A, B, C Unit 2 A, B, C Unit 3 A, B, C Unit 4 A, B, C | CO6: To obtain Practical Experience Working with all real data setTo introduce students to basic applications, concepts, and techniqueTo develop skills for using recent data mining software to solvein a variety of disciplines. To gain experience doing independent setISLab. Experiment 1Installation of WEKA ToolCreating new Arff FileLab. Experiment 2Pre-Processes Techniques on Data SetPre-process a given dataset based on Handling Missing ValuesLab. Experiment 3Generate Association Rules using the Apriori AlgorithmGenerating association rules using fp growth algorithmLab. Experiment 4Build a Decision Tree by using J48 algorithmNaïve bayes classification on a given data set | ues of data mining practical problem study and research CO Mapping CO1 CO2 |
| | Descriptio n Outline syllabu Unit 1 A, B, C Unit 2 A, B, C Unit 3 A, B, C Unit 4 | CO6: To obtain Practical Experience Working with all real data set To introduce students to basic applications, concepts, and techniqu To develop skills for using recent data mining software to solve in a variety of disciplines. To gain experience doing independent set Is Lab. Experiment 1 Installation of WEKA Tool Creating new Arff File Lab. Experiment 2 Pre-Processes Techniques on Data Set Pre-process a given dataset based on Handling Missing Values Lab. Experiment 3 Generate Association Rules using the Apriori Algorithm Generating association rules using fp growth algorithm Lab. Experiment 4 Build a Decision Tree by using J48 algorithm Naïve bayes classification on a given data set Lab. Experiment 5 | ues of data mining practical problem study and research CO Mapping CO1 CO2 CO2 |
| | Descriptio n Outline syllabu Unit 1 A, B, C Unit 2 A, B, C Unit 3 A, B, C Unit 4 A, B, C | CO6: To obtain Practical Experience Working with all real data set To introduce students to basic applications, concepts, and techniqu To develop skills for using recent data mining software to solve in a variety of disciplines. To gain experience doing independent set Lab. Experiment 1 Installation of WEKA Tool Creating new Arff File Lab. Experiment 2 Pre-Processes Techniques on Data Set Pre-process a given dataset based on Handling Missing Values Lab. Experiment 3 Generate Association Rules using the Apriori Algorithm Generating association rules using fp growth algorithm Lab. Experiment 4 Build a Decision Tree by using J48 algorithm Naïve bayes classification on a given data set Lab. Experiment 5 Applying k-means clustering on a given data set. Calculating Information gains measurs OLAP Cube and its different operations | ues of data mining practical problem study and research CO Mapping CO1 CO2 CO2 |
| | Descriptio n Outline syllabu Unit 1 A, B, C Unit 2 A, B, C Unit 3 A, B, C Unit 4 A, B, C Unit 4 A, B, C Unit 5 | CO6: To obtain Practical Experience Working with all real data set To introduce students to basic applications, concepts, and techniqu To develop skills for using recent data mining software to solve in a variety of disciplines. To gain experience doing independent set Is Lab. Experiment 1 Installation of WEKA Tool Creating new Arff File Lab. Experiment 2 Pre-Processes Techniques on Data Set Pre-process a given dataset based on Handling Missing Values Lab. Experiment 3 Generate Association Rules using the Apriori Algorithm Generating association rules using fp growth algorithm Lab. Experiment 4 Build a Decision Tree by using J48 algorithm Naïve bayes classification on a given data set Lab. Experiment 5 Applying k-means clustering on a given data set. Calculating Information gains measurs OLAP Cube and its | ues of data mining practical problem study and research CO Mapping CO1 CO2 CO3 CO4 |



| Text book/s* | 1. Jiawei Han and M Kamber, Data Mining Concepts and | |
|--------------|--|--|
| | Techniques, , Second Edition, Elsevier Publication. | |
| Other | 1. Arun K. Pujari, Data Mining Techniques, University Press. | |
| | 2. Vipin Kumar, Introduction to Data Mining Pang Ning Tan, | |
| S | Michael Steinbach, Addison Wesley. | |

| РО | РО | PO | РО | PO | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA270.1 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 1 | 2 | 1 | 2 | 2 |
| BDA270.2 | 1 | 2 | 3 | 2 | 2 | 1 | 1 | 3 | 1 | 1 | 2 | 1 | 2 | 2 |
| BDA270.3 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 1 | 2 | 1 | 2 | 2 |
| BDA270.4 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 1 | 2 | 1 | 2 | 2 |
| BDA270.5 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 1 | 2 | 1 | 2 | 2 |
| BDA270.6 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 1 | 2 | 1 | 2 | 2 |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | 2.0 | 1.0 | 1.0 | 3.0 | 1.0 | 1.0 | 2.0 | 1.0 | 2.0 | 2.0 |



| Scho | ol: SSBSR | Batch: 2024-28 | |
|--------------|--------------------------|--|--|
| Prog | gramme: B.Sc. | Academic Year: 2025-26 | |
| (Hor | | | |
| Bran Anal | ich: Data Science & | z Semester: IV | |
| Anai 1 | Course Code | BDA271 | |
| 2 | Course Title | | |
| | Credits | Database Management Systems Lab | |
| 3 | | 1 | |
| 4 | Contact Hours | 0-0-2 | |
| | (L-T-P) Course Status | CC | |
| 5 | | To make students familiar with the data structure & algorithms. The | concept of |
| 5 | Course Objective | data organizations, data structure operations; analysis of an algorithm Queues; Linked Lists; Sorting and Hashing; Graph. | ; Stacks and |
| 6 | Course Outcomes | CO1: Explain and illustrate the concepts of basic terminologies: eler organizations, data structure operations: insertion, deletion, travers K3, K4) CO2: Describe the analysis of an algorithm, asymptotic; notations trade-off. (K1, K2, K3) CO3: Describe Linear Search and Binary Search Techniques and complexity analysis. (K2, K3, K4) CO4: Describe ADT Stack and its operations: Algorithms and their analysis, Applications of Stacks; Types of Queue; Algorithms and their (K2, K3, K4) CO5: Describe the Singly-linked lists; trees; algorithms and analys K6) CO6: Describe and analyze the basic concepts of Sorting and Hash (K1,K2, K4) | al, etc. (K2, s, time-space explain their r complexity heir analysis. sis. (K2, K3, ing; Graphs. |
| 7 | Course | This course introduces data structure & algorithms. The conce | ept of data |
| | Description | organizations, data structure operations; analysis of an algorithm; | |
| | | Queues; Linked Lists; Sorting and Hashing; Graph. | |
| 8 | Outline syllabus | | CO Mapping |
| | Unit 1 | | |
| | A, B, C | Problem-based on uses functions to perform the following operations on a singly linked list i) Creation ii) Insertion iii) Deletion iv) Traversal. Problem-based on uses functions to perform the following operations on the doubly linked list i) Creation ii) Insertion iii) Deletion iv) Traversal. | CO1, CO2 |
| | Unit 2 | | |
| | A, B, C | Problem-based on uses functions to perform the following operations on circular linked List i) Creation ii) Insertion iii) Deletion iv) Traversal. Problem-based on implement stack (its operations) using i) Arrays ii) Linked list(Pointers). | CO1, CO3 |
| | Unit 3 | | |
| | A, B, C | Problem-based on implementing Queue (its operations) using i) Arrays ii) Linked list (Pointers). Problem-based on implementing Circular Queue using arrays. Problem-based on both recursive and nonrecursive functions to perform the following searching operations for a Key value in a given list of integers: a) Linear search b) Binary search. | CO1, CO4 |
| | | | |
| | Unit 4 | | |
| | Unit 4 A, B, C | Problem-based on implements the following sorting i) Bubble sort ii) Selection sort iii) Quick sort. Problem-based on implements the following i) Insertion sort ii) Merge sort iii) Heap sort. Problem- based on implementing all the functions of a dictionary (ADT) using Linked List. | CO1, CO5 |



| A, B, C | Problem-based on performing the following operations: a) Insert an element into a binary search tree. b) Delete an element from a binary search tree. c) Search for a key element in a binary search tree. Problem-based on to implement the tree traversal methods. Problem-based on performing the following operations: a) Insert an element into an AVL tree. b) Delete an element from an AVL tree. c) Search for a key element in an AVL tree. | CO1, CO5, CO6 |
|---------------------------|--|------------------|
| Mode of examination | Practical+Viva | |
| Weightage Distribution | CA:25%; CE:25%; ESE:50% | |
| Text book/s* | 1. Fundamentals of Data Structures", Illustrated Edition by Ellis Horowitz, SartajSahni, Computer Science Press. | |
| Other References | Algorithms, Data Structures, and Problem-Solving with C++", Illustrated Edition by Mark Allen Weiss, Addison-Wesley Publishing Company. How to Solve it by Computer", 2nd Impression by R. G. Dromey, Pearson Education. | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA271.1 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | | 2 | |
| BDA271.2 | 1 | 2 | 3 | 2 | | 1 | 1 | 3 | 1 | | 2 | | 2 | |
| BDA271.3 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | | 2 | |
| BDA271.4 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | | 2 | |
| BDA271.5 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | | 2 | |
| BDA271.6 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 2 | | 2 | |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 2.0 | | 2.0 | |



| Scho | ol: SSBSR | Batch: 2024-28 | |
|------|-------------------|--|----------------|
| Prog | gramme: B.Sc. | Academic Year: 2025-26 | |
| (Hor | | | |
| | ich: Data Science | Semester: IV | |
| | nalytics | DD 4 272 | |
| 1 | Course Code | BDA272 | |
| 2 | Course Title | Sampling Theory Lab | |
| 3 | Credits | | |
| 4 | Contact Hours | 0-0-2 | |
| | (L-T-P) | | |
| | Course Status | DSE | |
| 5 | Course | This course initiates the advanced concept of sample and population enumeration versus sampling. The concept of Systematic Sampling, | on, complete |
| | Objective | the population mean and total, variances of these estimates along with | h the brief of |
| | | the present official statistical system in India, methods of collection | n of official |
| - | <u> </u> | statistics, their reliability, and limitations have been introduced. | |
| 6 | Course | CO1: Explain and illustrate the concepts of sample and population. (K) CO2: Describe the properties of complete enumeration versus sample | 2, K3, K4) |
| | Outcomes | random sampling with and without replacement. (K1, K2, K3) | |
| | | CO3: Describe estimates of the population mean, explain its appl | ication and |
| | | estimates of these variances, and sample size determination. (K2, K3, I CO4: Describe stratified random sampling, estimates of the population | n mean and |
| | | total and explain its application, and illustrate systematic sampling. (K | 2, K3, K4). |
| | | CO5: Describe the ratio and regression methods of estimation a variances in terms of the correlation coefficient between X and | nd evaluate |
| | | regression method and their comparison with SRS. (K2, K3, K6). | Y for the |
| | | CO6: Describe and analyze the basic concepts present official statistic India, and methods of collection of official statistics. (K1,K2, K4). | al system in |
| | | | |
| 7 | Course | This is an advanced course in statistics. Students are introduced to the | |
| | Description | involved in using sample data to make inferences about populations. I | |
| | | the study of measures of central tendency and dispersion, finite statistical inferences from large and small samples, linear regre | |
| | | correlation and hypothesis. | |
| 8 | Outline syllabus | | СО |
| | • | | Mapping |
| | Unit 1 | Lab. Experiment 1 | |
| | A, B, C | Problem based on how to draw the sample from the population in SRSWR and SRSWOR | CO1, CO2 |
| | Unit 2 | Lab. Experiment 2 | |
| | A, B, C | Problem-based on simple random sampling and find that SRSWOR | CO1, CO3 |
| | | performs better than SRSWR | |
| | Unit 3 | Lab. Experiment 3 | |
| | A, B, C | Problem-based on stratified random sampling | CO1, CO4 |
| _ | Unit 4 | Lab. Experiment 4 | |
| | A, B, C | Problem-based on systematic sampling | CO1,CO5 |
| | Unit 5 | Lab. Experiment 5 | |
| | A, B, C | Problem-based on ratio and regression type estimator. | CO1, CO6 |
| | Mode of | Practical+Viva | |
| | examination | | |
| | Weightage | CA: 25%; CE:25%; ETE:75% | |
| | Distribution | CA: 25%; CE:25%; ETE:75% | |
| | | 1 Martha MNI (1077). Samuling Theory 9. Statistical Mathedra | |
| | Text book/s* | 1.Murthy M.N. (1977): Sampling Theory & Statistical Methods, | |
| | Text book/s* | Statistical Pub. Society, Calcutta | |
| | Text book/s* | | |



| Other | 1. Mukhopadhyay P. (1998): Theory and Methods of Survey |
|------------|---|
| References | Sampling, Prentice Hall |
| | 2. Guide to current Indian Official Statistics, Central Statistical Organization, GOI, New Delhi. |

| РО | PO | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA272.1 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | | 2 | 1 | | |
| BDA272.2 | 1 | 2 | 3 | 2 | 1 | 1 | 1 | 3 | 1 | | 2 | 1 | | |
| BDA272.3 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | | 2 | 1 | | |
| BDA272.4 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | | 2 | 1 | | |
| BDA272.5 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | | 2 | 1 | | |
| BDA272.6 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | | 2 | 1 | | |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | 1.0 | 1.0 | 1.0 | 3.0 | 1.0 | | 2.0 | 1.0 | | |



Detailed Syllabus for

DEGREE IN

DATA SCIENCE & ANALYTICS



| | ool: SSBSR | Batch: 2024-28 | |
|------|--|--|---|
| (Hoi | | Academic Year: 2026-27 | |
| | nch: Data Science nalytics | | |
| 1 | Course Code | BDA346 | |
| 2 | Course Title | Artificial Intelligence | |
| 3 | Credits | 5 | |
| 4 | Contact Hours (L-T-P) | 5-0-0 | |
| | Course Status | СС | |
| 5 | Course Objective | The objective of this course is to help students to learn the application learning /A. I have algorithms in the different fields of science, med etc. | |
| 6 | Course Outcomes | CO1: Understand basic concepts and applications of machine learningK4).CO2: Able to predicate logic and transform real-life information int | |
| | | representation. (K3, K6). CO3: Analyze the state space and its searching strategies. (K2, K5). CO4: Able to apply machine learning concepts and a range of probl be handled by machine learning. (K2, K3, K4). CO5: Analyze problem specifications and derive appropriate solution for them and also design and implement appropriate solutions problems and planning problems. (K4, K6). CO6: Enable students to apply the machine learning concepts problems. (K5, K6) | n techniques s for search |
| 7 | Course | This serves since to introduce the fear demonstral serves at A stifficial | |
| | Description | This course aims to introduce the fundamental concepts of Artificial to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. | as searching |
| 8 | Description Outline syllabus | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. | as searching |
| - | Description | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. | as searching al Language CO Mapping |
| - | Description Outline syllabus | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. | as searching al Language CO Mapping CO1 |
| - | Description Outline syllabus Unit 1 | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. | as searching al Language CO Mapping |
| - | Description Outline syllabus Unit 1 A | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. Overview of AI problems, AI problems as NP, NP-Complete, and NP-Hard problems. Strong and weak, neat and scruffy, symbolic and sub-symbolic, knowledge-based and data-driven AI. Search Strategies: Problem spaces (states, goals, and operators), | as searching al Language CO Mapping CO1 |
| - | Description Outline syllabus Unit 1 A B | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. Overview of AI problems, AI problems as NP, NP-Complete, and NP-Hard problems. Strong and weak, neat and scruffy, symbolic and sub-symbolic, knowledge-based and data-driven AI. | as searching al Language CO Mapping CO1 CO1 |
| - | Description Outline syllabus Unit 1 A B C | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. Overview of AI problems, AI problems as NP, NP-Complete, and NP-Hard problems. Strong and weak, neat and scruffy, symbolic and sub-symbolic, knowledge-based and data-driven AI. Search Strategies: Problem spaces (states, goals, and operators), problem-solving by search, Heuristics, and informed search, Minmax Search, Alpha-beta pruning. Constraint satisfaction (backtracking and local search methods). | as searching al Language CO Mapping CO1 CO1 CO1 |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B B | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. Overview of AI problems, AI problems as NP, NP-Complete, and NP-Hard problems. Strong and weak, neat and scruffy, symbolic and sub-symbolic, knowledge-based and data-driven AI. Search Strategies: Problem spaces (states, goals, and operators), problem-solving by search, Heuristics, and informed search, Minmax Search, Alpha-beta pruning. Constraint satisfaction (backtracking and local search methods). Knowledge representation and reasoning: propositional and predicate logic, Resolution and theorem proving Temporal and spatial reasoning. | as searching al Language CO Mapping CO1 CO1 CO1 CO1 CO2 |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C C | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. Overview of AI problems, AI problems as NP, NP-Complete, and NP-Hard problems. Strong and weak, neat and scruffy, symbolic and sub-symbolic, knowledge-based and data-driven AI. Search Strategies: Problem spaces (states, goals, and operators), problem-solving by search, Heuristics, and informed search, Minmax Search, Alpha-beta pruning. Constraint satisfaction (backtracking and local search methods). Knowledge representation and reasoning: propositional and predicate logic, Resolution and theorem proving Temporal and | as searching al Language CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C C Unit 3 | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. Overview of AI problems, AI problems as NP, NP-Complete, and NP-Hard problems. Strong and weak, neat and scruffy, symbolic and sub-symbolic, knowledge-based and data-driven AI. Search Strategies: Problem spaces (states, goals, and operators), problem-solving by search, Heuristics, and informed search, Minmax Search, Alpha-beta pruning. Constraint satisfaction (backtracking and local search methods). Knowledge representation and reasoning: propositional and predicate logic, Resolution and theorem proving Temporal and spatial reasoning. Probabilistic reasoning, Bayes theorem. | as searching al Language CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO2 |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C C | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. Overview of AI problems, AI problems as NP, NP-Complete, and NP-Hard problems. Strong and weak, neat and scruffy, symbolic and sub-symbolic, knowledge-based and data-driven AI. Search Strategies: Problem spaces (states, goals, and operators), problem-solving by search, Heuristics, and informed search, Minmax Search, Alpha-beta pruning. Constraint satisfaction (backtracking and local search methods). Knowledge representation and reasoning: propositional and predicate logic, Resolution and theorem proving Temporal and spatial reasoning. Probabilistic reasoning, Bayes theorem. Totally-ordered and partially-ordered Planning. | as searching al Language CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C C Unit 3 | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. Overview of AI problems, AI problems as NP, NP-Complete, and NP-Hard problems. Strong and weak, neat and scruffy, symbolic and sub-symbolic, knowledge-based and data-driven AI. Search Strategies: Problem spaces (states, goals, and operators), problem-solving by search, Heuristics, and informed search, Minmax Search, Alpha-beta pruning. Constraint satisfaction (backtracking and local search methods). Knowledge representation and reasoning: propositional and predicate logic, Resolution and theorem proving Temporal and spatial reasoning. Probabilistic reasoning, Bayes theorem. Totally-ordered and partially-ordered Planning. Goal stack planning, Nonlinear planning, Hierarchical planning. | as searching al Language CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO2 |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C C Unit 2 A B C Unit 3 A | to students. The course will explain various important concepts such techniques, Knowledge representation, Uncertainty, and Natura Processing. Overview of AI problems, AI problems as NP, NP-Complete, and NP-Hard problems. Strong and weak, neat and scruffy, symbolic and sub-symbolic, knowledge-based and data-driven AI. Search Strategies: Problem spaces (states, goals, and operators), problem-solving by search, Heuristics, and informed search, Minmax Search, Alpha-beta pruning. Constraint satisfaction (backtracking and local search methods). Knowledge representation and reasoning: propositional and predicate logic, Resolution and theorem proving Temporal and spatial reasoning. Probabilistic reasoning, Bayes theorem. Totally-ordered and partially-ordered Planning. Goal stack planning, Nonlinear planning, Hierarchical | as searching al Language CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO2 CO2 |



| A | Natural Language Processing: Language models, n-grams, Vector space models, Bag of words, Text classification. Information retrieval. | CO5 |
|--------------|---|-----|
| В | Agents: Definition of agents, Agent architectures (e.g., reactive, layered, cognitive). | CO5 |
| С | Multi-agent systems-Collaborating agents, Competitive agents, Swarm systems and biologically inspired models. | |
| Unit 5 | Intelligent Systems: Representing and Using Domain Knowledge, Expert System Shells, Explanation, Knowledge Acquisition. | CO6 |
| А | Key Application Areas: Expert system, decision support systems. | CO6 |
| В | Speech and vision, Natural language processing, Information Retrieval, Semantic Web. | CO6 |
| С | Area of parallelogram and quadrilateral, Vector triple product. | CO6 |
| Mode of | Theory | |
| examination | | |
| Weightage | CA:25%; ESE:75% | |
| Distribution | CA.2570, ESE.7570 | |
| Text book/s* | Artificial Intelligence Elaine Rich, Kevin Knight, and Shivashankar B Nair, Tata McGraw Hill. | |
| Other | 1. Introduction to Artificial Intelligence and Expert Systems by Dan | |
| References | W. Patterson, Pearson Education. | |
| | 2. Artificial Intelligence: A Modern Approach by S. Russell and P. | |
| | Norvig, Prentice Hall. | |

| РО | РО | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA346.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | 3 | | |
| BDA346.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 3 | 3 | | |
| BDA346.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 3 | 3 | | |
| BDA346.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 3 | 3 | | |
| BDA346.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | 3 | | |
| BDA346.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 3 | 3 | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 3.0 | 3.0 | | |



| | ool: SSBSR | Batch: 2024-28 | |
|------------|-----------------------------|---|-----------------|
| | gramme: B.Sc. | Academic Year: 2026-27 | |
| (Hor | | | |
| | ich: Data Science | Semester: V | |
| a A | nalytics Course Code | BDA303 | |
| 2 | Course Code Course Title | Machine learning | |
| 2 3 | Course Title Credits | | |
| | | 4 | |
| 4 | Contact Hours | 4-0-0 | |
| | (L-T-P) | | |
| | Course Status | СС | |
| 5 | Course | The objective of this course is to introduce machine learning fundament | ntals to |
| | Objective | students. | |
| 6 | Course | CO1: Recognize the characteristics of machine learning that make it u | seful to real- |
| | Outcomes | world problems (K2, K3) | |
| | | CO2: Characterize machine learning algorithms as supervised, sen | ni-supervised, |
| | | and unsupervised (K2, K3) | |
| | | CO3: Design and implement machine learning solutions to | classification, |
| | | regression, and clustering problems (K3, K6). | |
| | | CO4: Be able to evaluate and interpret the results of the algorithms (K | 4, K5) |
| | | CO5: Effectively use machine learning toolboxes (K5). | |
| | | CO6: Ability to recognize and implement various ways of selecting s | |
| | | parameters for different machine learning techniques. Ability to i | ntegrate deep |
| | | learning libraries and mathematical and statistical tools (K4, K5). | |
| 7 | Course | This course provides introductory concepts of various machin | |
| | Description | techniques to students which will help to build the foundation | |
| | | understanding. This course also aims to provide details of various ste | |
| | | in the machine learning pipeline such as data collection, pre-process | |
| | | engineering, etc. This course also introduces popular tools used in | |
| | | machine learning. This course mainly focused on Regression and Neu | ral network- |
| 0 | | based Machine learning algorithms. | СО |
| 8 | Outline syllabus | | Mapping |
| | Unit 1 | Introduction to Machine Learning | mapping |
| | A | Machine Learning Fundamentals –Types of Machine Learning - | CO1 |
| | 1 | Supervised, Unsupervised, Reinforcement- The Machine Learning | 001 |
| | | process. | |
| | | Terminologies in ML- Testing ML algorithms: Over fitting, | CO1 |
| | В | Training, Testing and Validation Sets-Confusion matrix -Accuracy | 001 |
| | D | metrics- ROC Curve. | |
| | С | Basic Statistics: Averages, Variance and Covariance, The Gaussian- | CO1 |
| | | The Bias-Variance trade off- Applications of Machine Learning. | 001 |
| | Unit 2 | | |
| | A | Regression: Linear Regression – Multivariate Regression analysis, | CO2 |
| | | Linear Basis Function Models, The Bias-Variance Decomposition, | |
| | | Bayesian Linear Regression | |
| | В | Classification: Linear Discriminant Analysis, Logistic Regression- | CO2 |
| | | K-Nearest Neighbor classifier. | ~~ = |
| | С | Decision Tree based methods for classification and Regression- | CO2 |
| | | Ensemble methods. | 202 |
| | Unit 3 | | |
| | A | Clustering- K-Means clustering, Hierarchical clustering. | CO3 |
| | B | The Curse of Dimensionality –Dimensionality Reduction - Principal | CO3 |
| | u u | Component Analysis - Probabilistic PCA- Independent Components | 005 |
| | | | |



| | analysis | |
|---------------------------|--|-----|
| С | The Internet, Business and Retail, Law Enforcement, Computing, Clustering models: How the K-means and PCA works, Calculating the number of clusters in a dataset. | CO3 |
| Unit 4 | | |
| А | Perceptron- Multilayer perceptron- Back Propagation- Initialization, Training and Validation Support. | CO4 |
| В | Vector Machines (SVM) as a linear and non-linear classifier - Limitations of SVM | CO4 |
| С | Recognition of MNIST handwritten digits using Artificial Neural Network. Build an email spam classifier using SVM. | CO4 |
| Unit 5 | | |
| А | Bayesian Networks - Learning Naive Bayes classifiers-Markov Models – Hidden Markov Models. | CO5 |
| В | Sampling – Basic sampling methods – Monte Carlo -Reinforcement Learning. | CO5 |
| С | Classify the given text segment as 'Positive' or 'Negative' statement using the Naive Bayes Classifier. Predict future stock price of a company using Monte Carlo Simulation. | CO6 |
| Mode of examination | Theory | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | Mitchell Tom, Machine Learning. McGraw Hill. Dr. Nilesh Shelke, Dr. Gopal Sakarkar, Dr N V Choudhari, Introduction to Machine Learning, Ganu Prakashan. | |
| Other References | Trevor Hastie, Robert Tibshirani, Jerome Friedman, The Elements of Statistical Learning Data Mining, Inference, and Prediction Andreas C. Muller, "Introduction to Machine Learning with Python: A Guide for Data Scientists", O'Reilly | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | PO | PO | PO | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA303.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | 3 | | |
| BDA303.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 3 | 3 | | |
| BDA303.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 3 | 3 | | |
| BDA303.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 3 | 3 | | |
| BDA303.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | 3 | | |
| BDA303.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 3 | 3 | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 3.0 | 3.0 | | |



| Scho | ool: SSBSR | Batch: 2024-28 | | | | | | | | |
|------|-------------------|--|--|--|--|--|--|--|--|--|
| Prog | gramme: B.Sc. | Academic Year: 2026-27 | | | | | | | | |
| (Hor | | | | | | | | | | |
| | ich: Data Science | Semester: V | | | | | | | | |
| | nalytics | DD 4 210 | | | | | | | | |
| 1 | Course Code | BDA319 | | | | | | | | |
| 2 | Course Title | Regression Analysis | | | | | | | | |
| 3 | Credits | 3 | | | | | | | | |
| 4 | Contact Hours | 3-0-0 | | | | | | | | |
| | (L-T-P) | | | | | | | | | |
| | Course Status | CC | | | | | | | | |
| 5 | Course | The main objective of this course is to demonstrate and intended to | | | | | | | | |
| | Objective | in the techniques necessary to understand and carry out regression | and predictive | | | | | | | |
| | | analysis. | | | | | | | | |
| 6 | Course | At the end of the course, the student should be able to | | | | | | | | |
| | Outcomes | CO1: Explain the concept of regression with two and multiple variab | oles. | | | | | | | |
| | | CO2: Testing of the single and subset of the regression coefficient. | | | | | | | | |
| | | CO3: Explain the concept of multicollinearity. CO4: Describe how to overcome the problem of heteroscedasticity a | nd | | | | | | | |
| | | autocorrelation. | liu | | | | | | | |
| | | CO5: Explain the concept of dummy variables. | | | | | | | | |
| | | CO6: How to apply logistic regression on a dataset. | | | | | | | | |
| 7 | Course | A PG-level course in regression analysis, intended to verse s | | | | | | | | |
| | Description | techniques necessary to understand and carry out methods of res analysis. Lectures study the large-sample properties of estimators sample, k-sample, and partial likelihood inference, with proofs counting process and Martingale theory. The theory of competing from several angles. Many extensions of the Cox model to more structures are considered. | earch in serial based on one- based on the risks is studied | | | | | | | |
| 8 | | | | | | | | | | |
| | Unit 1 | | | | | | | | | |
| | A | Simple Linear Regression: Simple linear regression model. Least- squares estimation of parameters. Hypothesis testing on the slope and intercept. Interval estimation in simple linear regression. | CO1 | | | | | | | |
| | В | Prediction of new observations. Coefficient of determination. Estimation by maximum likelihood. | CO1 | | | | | | | |
| | C | Multiple linear regression: Multiple linear regression models. | CO1 | | | | | | | |
| | | Estimation of the model parameters. Hypothesis testing in multiple linear regression. Confidence intervals in multiple regression. | | | | | | | | |
| | | Coefficient of determination and Adjusted R2. | | | | | | | | |
| | Unit 2 | | CO2 | | | | | | | |
| | A | Logistic Regression: Introduction, Linear predictor and link | CO2 | | | | | | | |
| | 1 | functions, logit, probit, odds ratio, the test of hypothesis. Discriminant Analysis. | 002 | | | | | | | |
| | В | Model Adequacy: Checking of linearity between study and explanatory variable, Residual Analysis, Detection and treatment of outliers, Residual plots. | CO2 | | | | | | | |
| | С | The PRESS statistic. Outlier test based on Studentized Residual (R-student). Test for lack of fit of the regression model. | | | | | | | | |
| | Unit 3 | | CO3 | | | | | | | |
| | А | Data Understanding and Preparation Introduction, Reading data from various sources, Data visualization, Distributions, and | CO3 | | | | | | | |



| В | The extent of Missing Data. Segmentation, Outlier detection, | CO3 |
|--------------|--|----------|
| | Automated Data Preparation | |
| C | Combining data files, Aggregate Data, Duplicate Removal, | |
| | Sampling DATA, Data Caching, Partitioning data, and Missing | |
| | Values. | |
| Unit 4 | | CO4 |
| А | Model development & techniques Data Partitioning, Model | CO4 |
| | selection, Model Development Techniques | |
| В | Neural networks, Decision trees, Logistic regression, Discriminant | CO4 |
| | analysis, Support vector machine | |
| С | Bayesian Networks, Linear Regression, Cox Regression, and | |
| | Association rules. | |
| Unit 5 | | CO5 |
| А | Model Evaluation and Deployment Introduction, Model | CO5 |
| | Validation, Rule Induction Using CHAID | |
| В | Automating Models for Categorical and Continuous targets, | CO5, CO6 |
| | Comparing and Combining Models, and Evaluation Charts for | |
| | Model Comparison | |
| С | Meta Level Modeling, Deploying Model, Assessing Model | |
| | Performance, Updating a Model. | |
| Mode of | Theory | |
| examination | | |
| Weightage | | |
| Distribution | CA:25%; ESE:75% | |
| Text book/s* | 1. Johnston, J. (1984). Econometric Methods, McGraw Hill | |
| TCAT DOOK/S | Kogakusha Ltd. | |
| | | |
| Other | 1. Draper, N. R., and Smith, H. (1998). Applied Regression | |
| References | Analysis (John Wiley) Third edition. | |
| | | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | РО | РО | PO | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|----|----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA319.1 | 3 | 3 | 2 | 2 | 1 | 1 | | | | | 3 | 3 | | |
| BDA319.2 | 2 | 3 | 3 | 2 | 1 | 1 | | | | | 3 | 3 | | |
| BDA319.3 | 2 | 2 | 2 | 3 | 1 | 1 | | | | | 3 | 3 | | |
| BDA319.4 | 2 | 3 | 2 | 2 | 1 | 1 | | | | | 3 | 3 | | |
| BDA319.5 | 3 | 3 | 2 | 2 | 1 | 1 | | | | | 3 | 3 | | |
| BDA319.6 | 3 | 3 | 2 | 3 | 1 | 1 | | | | | 3 | 3 | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | 1.0 | 1.0 | | | | | 3.0 | 3.0 | | |



| Scho | ool: SSBSR | Batch: 2024-28 | |
|-------------|-------------------------------|--|--------------------------------|
| | gramme: B.Sc. | Academic Year: 2026-27 | |
| <u>(Hor</u> | | | |
| | nch: Data Science nalytics | Semester: V | |
| 1 1 | Course Code | BDA320 | |
| 2 | Course Title | Advanced Statistical Analysis | |
| 3 | Credits | | |
| 4 | Contact Hours | | |
| 4 | (L-T-P) | 2-0-0 | |
| | Course Status | DSE | |
| 5 | Course | | anagialist to |
| 3 | | After completing this course, students are expected to become a | • |
| | Objective | analyze the observed phenomena at in advanced statistical l | |
| | | importantly, students are expected to provide an analytical solution t | - |
| | | using appropriately selected models and data and discover meaningfu from the solution. | l knowledge |
| 6 | 0 | | $(\mathbf{V}_1, \mathbf{V}_2)$ |
| 6 | Course Outcomes | CO1: Describe how to Differentiate various probability distributions. $(CO2)$ Understand the concern of estimation $(K2, K2)$ | (K1, K2) |
| | Outcomes | CO2: Understand the concept of estimation. (K2, K3) CO3: Know how to recognize the sampling distributions. (K2, K3) | |
| | | CO4: Learn non-parametric tests such as the chi-Square test for Indep | endence as |
| | | well as Goodness of Fit. (K3, K4) | endence us |
| | | CO5: Know how to apply various statistics and analyses. (K3, K4, K5 |) |
| | | CO6: Able to know statistical technique implantation in a practical si | |
| | | K4, K5) | |
| 7 | Course | This course provides students with the statistical foundation of | |
| | Description | problems of real life. Students will learn to recognize the main fea | |
| | | processes under investigation that could be analyzed in terms of statistical approaches. Grading this course will help the future a | |
| | | analyze the observed phenomena in advanced statistical level. | specialist to |
| 8 | | | |
| - | Unit 1 | | |
| | A | Use of discrete distribution (Uniform, Binomial, and Poisson) in real-life problems. | CO1, CO6 |
| | В | Use of continuous distribution (Normal, Exponential, and Gamma) in real-life problems. | CO1, CO6 |
| | С | Its applications in Industrial work. | CO1, CO6 |
| | Unit 2 | | |
| | А | Sampling Distributions. | CO2, CO6 |
| | В | χ^2 distribution properties and Interrelationships. | CO2, CO6 |
| | С | t distribution properties and Interrelationships. | CO2, CO6 |
| | Unit 3 | | |
| | А | F distribution properties. | CO3, CO6 |
| | В | Interrelationship of χ^2 , t, F distributions. | CO3, CO6 |
| | С | Point Estimation, Interval estimation for mean, the variance of normal population, and proportion of the binomial population. | CO3, CO6 |
| | Unit 4 | | |
| | А | Type I and Type II errors, Critical Region, Size of the test, P value, Power. | CO4, CO6 |
| | В | Large Sample test -Z test. | CO4, CO6 |
| | С | Large Sample test - Chi-Square test-goodness of fit, the test of | CO4, CO6 |



| Unit 5 | | |
|---------------------------|---|----------|
| А | ANOVA, | CO5, CO6 |
| В | Cluster and Principal Components Analysis (PCA). | CO5, CO6 |
| С | Factor Analysis, Canonical Correlation | CO5, CO6 |
| Mode of examination | Practical Based | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | Gupta. S.C. & Kapoor, V.K. (2002): Fundamentals of Mathematical Statistics, Sultan Chand & Sons. Westfall, P., & Henning, K. S. (2013): Understanding Advanced Statistical Methods. CRC Press. | |
| Other References | Croxton, Fredrick E., Cowden, Dudley J. and Klein, S. (1973): Applied General Statistics, 3rd Edition. Prentice Hall Mukhopadhyay P. (1999): Applied Statistics, Books and | |
| | Allied Pvt. Ltd. | |

| PO | РО | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|-----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA320.1 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 1 | |
| BDA320.2 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 1 | |
| BDA320.3 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 1 | |
| BDA320.4 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 1 | |
| BDA320.5 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 1 | |
| BDA320.6 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 1 | |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 3.0 | 3.0 | 1.0 | |



| Sch | ool: SSBSR | Batch: 2024-28 | |
|-----------------|--------------------------|---|---|
| Prog | gramme: B.Sc. | Academic Year: 2026-27 | |
| (Ho | | | |
| | nch: Data Science | Semester: V | |
| <u>а</u> А 1 | nalytics Course Code | BDA321 | |
| - | Course Title | | |
| 2 3 | Credits | Experimental Design | |
| | | | |
| 4 | Contact Hours (L-T-P) | 2-0-0 | |
| | Course Status | DSE | |
| 5 | Course Objective | The course objective is to learn how to plan, design and conduct efficiently and effectively, and analyze the resulting data to obta conclusions. | • |
| 6 | Course | After the completion of this course, the student will be able to | |
| | Outcomes | CO1: Build knowledge of basic principles of design of | |
| | | experiment. CO2: Make use of the concept to various simple types of expedesigns. CO3: Make use of the concept to f complex types of experimental des CO4: Evaluate the factorial experiment, confounding and split/strip pl CO5: Apply concept of missing-plot techniques, cross-over transformation of data and response question. CO6: How to design and conduct experiments, and how to analyze the answer various research questions | igns. ot design. design, and n properly to |
| 7 | Course Description | The course objective is to learn how to plan, design and conduct efficiently and effectively, and analyze the resulting data to obta conclusions. | |
| 8 | | | |
| | Unit 1 | | |
| | Α | Analysis of variance, | CO1 |
| | В | Basic principles of design of experiments. | CO1 |
| | С | Uniformity trials. | CO1 |
| | Unit 2 | | |
| | А | Completely randomized design (CRD), | CO2 |
| | В | Randomized complete block design (RCBD), | CO2 |
| | С | Latin square design (LSD) | CO2 |
| | Unit 3 | | |
| | А | Balanced incomplete block (BIB) design, | CO3 |
| | В | Resolvable block designs and their applications | CO3 |
| | С | Randomization procedure, analysis and interpretation of results. | CO3 |
| | Unit 4 | | |
| | А | Factorial experiments, | CO4 |
| | В | Confounding in factorial experiments-application in 2n and 3n factorial experiments. | CO4 |
| | С | Factorial experiments with extra treatment(s). Split plot and Strip plot designs | CO4 |
| | Unit 5 | | |
| | А | Groups of experiments. Analysis of covariance. | CO5 |



| В | Missing plot technique and its application to RCBD, LSD. Cross- over design. Sampling in field experiments. | CO5 |
|---------------------|--|-----|
| С | Transformation of data. Response surfaces. Experiments with mixtures. | CO6 |
| Mode of examination | Theory | |
| | | |
| Weightage | CA:25%; ESE:75% | |
| Distribution | CR.2570, ESE.7570 | |
| Text book/s* | 1. Gupta. S.C. & Kapoor, V.K.: Fundamentals of Mathematical | |
| | Statistics, Sultan Chand & Sons Pvt. Ltd. New Delhi. | |
| | 2. Westfall, P., & Henning, K. S.: Understanding Advanced Statistical Methods. CRC Press. | |
| Other | 1.Cochran, Wigand Cox, G.M. Experimental Designs. John Wiley | |
| References | and Sons. | |
| | 2.Das, M.N. and Giri, Design and Analysis of Experiments. New | |
| | Age International. | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | PO | РО | PO | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|-----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA321.1 | | 2 | 1 | 2 | | 1 | | 3 | | | 1 | 1 | | |
| BDA321.2 | | 2 | 1 | 2 | | 1 | | 3 | | | 1 | 1 | | |
| BDA321.3 | | 2 | 1 | 2 | | 1 | | 3 | | | 1 | 1 | | |
| BDA321.4 | | 2 | 1 | 2 | | 1 | | 3 | | | 1 | 1 | | |
| BDA321.5 | | 2 | 1 | 2 | | 1 | | 3 | | | 1 | 1 | | |
| BDA321.6 | | 2 | 1 | 2 | | 1 | | 3 | | | 1 | 1 | | |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 1.0 | 1.0 | | |



| Sch | ool: SSBSR | Batch: 2024-28 | |
|------------|-------------------------|--|------------------|
| | gramme: B.Sc. | Academic Year: 2026-27 | |
| (Ho | | | |
| | nch: Data Science | Semester: V | |
| a A | nalytics Course Code | RBL003 | |
| 2 | Course Title | Research Based Learning-3 | |
| 3 | Credits | 1 | |
| 4 | Contact Hours | 0-0-2 | |
| 4 | (L-T-P) | 0-0-2 | |
| | Course Status | Project | |
| 5 | Course | 1. Deep knowledge of a specific area of specialization. | |
| 5 | Objective | 2. Develop communication skills, especially in project wr | iting and oral |
| | Objective | presentation. Develop some time management skills. | und orun |
| 6 | Course | CO1: Explain the concept of research within the subject, as regard | s approaching a |
| | Outcomes | question, collecting and analyzing background material, and pres | |
| | | questions and conclusions. (K2, K4) | - |
| | | CO2: Construct and develop a deeper interest in mathematics | and a taste for |
| | | research. (K5, K6) | 1 1 (177.4 |
| | | CO3: Select and recommend activities that support their profession | onal goals. (K4, |
| | | K6) CO4: Develop effective project organizational skills. (K5) | |
| | | CO5: Analyse the problem and summarize research findings. (K4,K | 5) |
| | | CO6: Use research findings to develop education theory and practic | |
| 7 | Course | Maintain a core of mathematical and technical knowledge that is ad | |
| | Description | changing technologies and provides a solid foundation for future lea | |
| | | | _ |
| 8 | | | |
| | Unit 1 | Introduction | CO1 |
| | | | |
| | Unit 2 | Case study | CO1,CO2 |
| | | | |
| | Unit 3 | Conceptual | CO3, CO4 |
| | | | |
| | Unit 4 | Development | CO4, CO5 |
| | | | |
| | Unit 5 | Finalisation | CO5, CO6 |
| | Mode of | | |
| | examination | | |
| | Weightage | | 1 |
| | Distribution | | |
| | Text book/s* | | |
| | Other | | |
| | References | | |



| РО | PO | PO | РО | PO | PO | PO | PO | PO | РО | РО | PO | PSO | PSO | PSO |
|----------|----|-----|-----|-----|-----|-----|----|-----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| RBL003.1 | | 2 | 1 | 2 | 2 | 1 | | 3 | | | 2 | 2 | 2 | 2 |
| RBL003.2 | | 2 | 1 | 2 | 2 | 1 | | 3 | | | 2 | 2 | 2 | 2 |
| RBL003.3 | | 2 | 1 | 2 | 2 | 1 | | 3 | | | 2 | 2 | 2 | 2 |
| RBL003.4 | | 2 | 1 | 2 | 2 | 1 | | 3 | | | 2 | 2 | 2 | 2 |
| RBL003.5 | | 2 | 1 | 2 | 2 | 1 | | 3 | | | 2 | 2 | 2 | 2 |
| RBL003.6 | | 2 | 1 | 2 | 2 | 1 | | 3 | | | 2 | 2 | 2 | 2 |
| Average | | 2.0 | 1.0 | 2.0 | 2.0 | 1.0 | | 3.0 | | | 2.0 | 2.0 | 2.0 | 2.0 |



| Schoo | ol: SSBSR | Batch: 2024-28 | |
|-------|-------------------|--|----------------|
| | gramme: B.Sc. | Academic Year: 2026-27 | |
| (Hoi | ns.) | | |
| | nch: Data Science | Semester: V | |
| | nalytics | | |
| 1 | Course Code | INC001 | |
| 2 | Course Title | Industry Connect | |
| 3 | Credits | 2 | |
| 4 | Contact Hours | 0-0-4 | |
| | (L-T-P) | 0-0-4 | |
| | Course Status | Project | |
| 5 | Course | This course will expose students to applying theories learned in the cla | assroom and |
| | Objective | provides current technological developments relevant to the subject ar training. Students will be able to identify their career preferences and | rea of |
| | 5 | goals. | professional |
| 6 | Course | Students will be able to: | |
| - | Outcomes | CO1: Get familiar with industry principles and practices. | |
| | | CO2: Identify and analyze an appropriate problem. | |
| | | CO3: Develop teamwork and apply prior acquired knowledge in prob | lem-solving. |
| | | CO4: Demonstrate effective verbal and written communication skills. | |
| | | CO5: Practice scientists' responsibilities, self-understanding, self-d | iscipline, and |
| | | ethical standards. | |
| | | CO6: Identify the career preferences and professional goals. | |
| 7 | Course | The Internship aims to offer students the opportunity to apply their pr | ior acquired |
| / | Description | knowledge in problem-solving. Students will acquire skills importa | |
| | Desemption | management, discipline, self-learning, effective communication, and s | |
| 8 | | | |
| 0 | Unit 1 | | |
| | | Define chiesting and conditions for the internation energies | CO1 CO(|
| | A, B, C | Define objectives and conditions for the internship, ensuring students that it is related to the study path carried out at the | CO1,CO6 |
| | | University | |
| | Unit 2 | | |
| | A, B, C | Problem Definition and identification, Team/Group formation, and | CO2,CO6, |
| | А, Б, С | Project Assignment. Finalizing the problem statement, and | 002,000, |
| | | resource requirement, if any. | |
| | Unit 3 | | |
| | A, B, C | The internship work plan is drawn up by developing teamwork and | CO3,CO6, |
| | | applying prior acquired knowledge in problem-solving. | , |
| | Unit 4 | | |
| | A, B, C | Demonstrate and execute Project with the team. Submission of the | CO4,CO6 |
| | Unit 5 | evaluation form and final report completed by the intern. | |
| | | Final evaluation form completed by the supervisor at the Host | CO5,CO6 |
| | A, B, C | Organization and final presentation before the departmental | 003,006 |
| | | committee. | |
| | | | |
| | Mode of | | |
| | examination | | |
| | Weightage | | |
| | Distribution | | |
| | Text book/s* | | |
| | | | |
| | | | |



| Other | | | | | | | | | | | | | | |
|----------------------|---|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| References | | | | | | | | | | | | | | |
| COURSE OUTCOM | OURSE OUTCOMES – PROGRAMME OUTCOMES MAPPING TABLE | | | | | | | | | | | | | |
| | | | | | | | | | | | | PSO | PSO | |
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| INC001.1 | | 2 | 1 | 2 | 2 | 1 | 2 | 3 | 2 | 2 | 2 | 2 | 2 | 2 |
| INC001.2 | | 2 | 1 | 2 | 2 | 1 | 2 | 3 | 2 | 2 | 2 | 2 | 2 | 2 |
| INC001.3 | | 2 | 1 | 2 | 2 | 1 | 2 | 3 | 2 | 2 | 2 | 2 | 2 | 2 |
| INC001.4 | | 2 | 1 | 2 | 2 | 1 | 2 | 3 | 2 | 2 | 2 | 2 | 2 | 2 |
| INC001.5 | | 2 | 1 | 2 | 2 | 1 | 2 | 3 | 2 | 2 | 2 | 2 | 2 | 2 |
| INC001.6 | | 2 | 1 | 2 | 2 | 1 | 2 | 3 | 2 | 2 | 2 | 2 | 2 | 2 |
| Average | | 2.0 | 1.0 | 2.0 | 2.0 | 1.0 | 2.0 | 3.0 | 2.0 | 2.0 | 2.0 | 2.0 | 2.0 | 2.0 |



| Scho | ol: SSBSR | Batch: 2024-28 | |
|-----------------|--------------------------|--|--|
| Prog | ramme: B.Sc. | Academic Year: 2026-27 | |
| (Hon | | | |
| | ich: Data Science | Semester: V | |
| & Al | nalytics Course Code | BDA355 | |
| - | | | |
| 2 | Course Title | Machine Learning Lab | |
| 3 | Credits | 1 | |
| 4 | Contact Hours (L-T-P) | 0-0-2 | |
| | Course Status | CC | |
| 5 | Course Objective | Learn the basic concepts of Machine Learning algorithms. Make use of Data sets in implementing the machine learning algori Implement the machine learning concepts and algorithms in any suital of choice. | ble language |
| 6 | Course Outcomes | CO1: Show the implementation of linear and logistic Regression applications. CO2: Interpretation of existing models to understand the solution of CO3: Application of existing mathematical solutions to test real wor CO4: Analyse the logical ability to apply clustering approach hierarchical patterns existing in real life problems. CO5: Build the understanding of learning theory to glance the upco through it. CO6: Appraise recent trends in machine learning and applications | environment. Id problems. 1 to extract |
| 7 | Course Description | This course introduces computational learning paradigm for implementable understanding for supervised and unsupervised lea problem areas. | rning based |
| 8 | Outline syllabus | | CO Mapping |
| | Unit 1 | | CO1 |
| | A, B, C | Write a Program to load and view data set file. | CO1, CO6 |
| | | Write a program to implement simple linear regression using housing price prediction problem. | |
| | | Write a program to implement binary logistic regression using cancer identification problem. | |
| | Unit 2 | | |
| | A, B, C | Write a program to implement gradient descent method for learning. | CO1, CO2 |
| | | Write a program to implement regularized linear regression. | |
| | | Write a program to implement regularized logistic regression. | |
| | | Write a program to Normalize the data used in linear regression problem above before predicting prices, and then predict the housing prices. | |
| | Unit 3 | | |
| | A, B, C | Write a program to implement Support Vector Machine regression using suitable dataset. | CO1, CO2, CO6 |
| | | Build an Artificial Neural Network by implementing the Backpropagation algorithm and test the same using appropriate data sets. Write a program to demonstrate the working of the decision tree based ID3 algorithm. Use an appropriate data set for building the decision tree and apply this knowledge to classify a new sample. | |
| | | Write a program to demonstrate the working of the Random Forest algorithm. Use an appropriate data set for classifying a new sample. | |



| Unit 4 | | |
|---------------------------|---|------------------|
| A, B, C | Write a program to implement K-Means clustering algorithm using an appropriate dataset. | CO2, CO3, CO4 |
| | Write a program to implement K-Means clustering algorithm using an appropriate dataset | |
| Unit 5 | | |
| A, B, C | Write a program to implement data split into training, cross validation and testing data. | CO4, CO5, CO6 |
| | Implement an Ensemble approach by combining different models to solve time series based prediction problem. | |
| | Conduct hypothesis testing using some statistical toolkit on appropriate problem. | |
| Mode of examination | Practical+Viva | |
| Weightage Distribution | CA:25%; CE:25%; ESE:50% | |
| Text book/s* | Bishop, C.: Pattern Recognition and Machine Learning. Berlin: Springer-Verlag. Ethem Alpaydin: Introduction to Machine Learning, 3rd Edition | |
| Other | 1. Baldi, P. and Brunak, S. (2002). Bioinformatics: A Machine | |
| References | Learning Approach. Cambridge, MA: MIT Press. | |
| | 2. https://www.toptal.com/machine-learning/ensemble- | |
| | methodsmachine-learning. | |

| РО | РО | PO | РО | PO | PO | PO | PO | PO | PO | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA355.1 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 2 | 3 | 2 | 2 | 3 |
| BDA355.2 | 1 | 2 | 3 | 2 | 2 | 1 | 1 | 3 | 1 | 2 | 3 | 2 | 2 | 3 |
| BDA355.3 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 2 | 3 | 2 | 2 | 3 |
| BDA355.4 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 2 | 3 | 2 | 2 | 3 |
| BDA355.5 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 2 | 3 | 2 | 2 | 3 |
| BDA355.6 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 2 | 3 | 2 | 2 | 3 |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | 2.0 | 1.0 | 1.0 | 3.0 | 1.0 | 2.0 | 3.0 | 2.0 | 2.0 | 3.0 |



| | ool: SSBSR | Batch: 2024-28 | |
|-----------------|---|---|---|
| | gramme: B.Sc. | Academic Year: 2026-27 | |
| (Ho | | | |
| | nch: Data Science analytics | Semester: V | |
| a A 1 | Course Code | BDA356 | |
| 2 | Course Title | | |
| | | Regression Analysis Lab | |
| 3 | Credits | 1 | |
| 4 | Contact Hours (L-T-P) | 0-0-2 | |
| | Course Status | CC | |
| 5 | Course Objective | After studying these courses students will be able to understand he the power of the test, analyze the multivariate data and u characteristics of multivariate quantitative research, including weaknesses. It also discusses the principles and characteristics of t data analysis techniques. | nderstand the strengths and |
| 6 | Course Outcomes | At the end of the course, the student should be able to CO1: Estimate the parameter by MLE CO2: Learn about how to calculate the Rao, Lehman, and Bhattachan CO3: Learn how to calculate the critical region, power of the test, and Neyman structure. CO4: Understand the basic concepts of multivariate normal distribut CO5: Calculate Wishart distribution in the multivariate analysis also find Mahalanobis D2 and HottelingT2. CO6: Apply the classification rule, PCA, and factor analysis. | unbiased test, |
| 7 | Course Description Outline syllabus | In this course, students are concerned with making inferences base found in the sample, to relations in the population. Also multivari data deals with examining the interrelationship between three or important variables or explaining variation in, usually one (or n dependent variable(s) based on two or more independent (explaining | ate analysis of more equally tore than one) |
| 0 | • | | Mapping |
| | Unit 1 | | |
| | A, B, C | Problem-based on Multiple regression analysis python using R/Python. | CO1 CO2 |
| | Unit 2 | | |
| | A, B, C | Problem-based on Logistic regression analysis python using R/Python. | ,CO2, CO3 |
| | Unit 3 | | |
| | A, B, C | Problem-based on Discriminant Analysis using R/Python. | CO3, CO4 |
| | Unit 4 | | |
| | A, B, C | Problem-based on Multivariate Analysis of Variance and Covariance | CO4,CO5 |
| | | using R/Python. | |
| | Unit 5 | using R/Python. | |
| | Unit 5 A, B, C | Problem-based on classification rule, PCA, and factor analysis using R/Python. | ,CO5, CO6 |
| | | Problem-based on classification rule, PCA, and factor analysis using | ,CO5, CO6 |
| | A, B, C Mode of | Problem-based on classification rule, PCA, and factor analysis using R/Python. | ,CO5, CO6 |



| Other | 2. Draper, N. R., and Smith, H. (1998). Applied Regression | |
|------------|--|--|
| References | Analysis (John Wiley) Third edition. | |

| РО | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA356.1 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 2 | 2 | 2 | 2 | 3 |
| BDA356.2 | 1 | 2 | 3 | 2 | 2 | 1 | 1 | 3 | 1 | 2 | 2 | 2 | 2 | 3 |
| BDA356.3 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 2 | 2 | 2 | 2 | 3 |
| BDA356.4 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 2 | 2 | 2 | 2 | 3 |
| BDA356.5 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 2 | 2 | 2 | 2 | 3 |
| BDA356.6 | 1 | 2 | 2 | 2 | 2 | 1 | 1 | 3 | 1 | 2 | 2 | 2 | 2 | 3 |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | 2.0 | 1.0 | 1.0 | 3.0 | 1.0 | 2.0 | 2.0 | 2.0 | 2.0 | 3.0 |



| School: SSBSR Programme: B.Sc. (Hons.) Branch: Data Science & Analytics | | Batch: 2024-28 | |
|---|-----------------------------|--|------------------|
| | | Academic Year: 2026-27 | |
| | | Semester: V | |
| 1 | Course Code | BDA359 | |
| 2 | Course Title | Advanced Statistical Analysis Lab | |
| 3 | Credits | 1 | |
| 4 | Contact Hours(L-T- P) | 0-0-2 | |
| | Course Status | DSE | |
| 5 | Course Objectiv e | After completing this course, students are expected to become a specialist to analyze the observed phenomena at in advanced statistical level. More importantly, students are expected to provide an analytical solutions to a problem using appropriately selected models and data and discover meaningful knowledge from the solution. | |
| 6 | Course Outcome | CO1: Describe how to Differentiate various probability distributio | ns. (K1,K2) |
| | s | CO2: Understand the concept of estimation. (K2,K3) | |
| | | CO3: Know how to recognize the sampling distributions. (K2,K3) | |
| | | CO4: Learn non-parametric tests such as the chi-Square test for Indas well as Goodness of Fit. (K3,K4) | dependence |
| | | CO5: Know how to apply various statistics and analyses. (K3,K4,I CO6: Able to know statistical technique implantation in a practica (K3,K4,K5) | |
| 7 | Course Descriptio n | This course provides students with the statistical foundation of the various problems of real life. Students will learn to recognize the main features of the processes under investigation that could be analyzed in terms of advanced statistical approaches. Grading this course will help the future specialist to analyze the observed phenomena in advanced statistical level. | |
| 8 | Outline syllab | | CO Mapping |
| | Unit 1 | Lab. Experiment 1: | mapping |
| | A, B, C | Real life Problem Based on Discrete Probability Distributions (Uniform, Binomial and Poisson) Using Python Real life problem based on Continuous Probability Distributions (Normal, Exponential and Gamma) using python. | CO1, CO2 |
| | Unit 2 | Lab. Experiment 2: | |
| | A, B, C | Real life Problem based on sampling Distribution (t-test and z test) using python. Real life Problem based on Sampling Distribution (F, Chi-Square) using python. | CO2, CO3 |
| | Unit 3 | Lab. Experiment 3: | |
| | A, B, C | Real life Problem based on ANOVA (One Way) using python. Real life Problem based on ANOVA (Two Way) using python. | CO3, CO4 |
| | Unit 4 | Lab. Experiment 4: | |
| | A, B, C | Real life Problem Based on Principle Component Analysis (PCA) in Python. Real life Problem Based on Factor Analysis in Python. | CO4, CO5, CO6 |
| | Unit 5 | Lab. Experiment 5: | |
| | A, B, C | Real life Problem Based on Cluster in Python. Real life problem based on Point Estimation and Interval. Estimation using Python | CO5, CO6 |



| Mode of examinatio n | Practical+Viva | |
|-------------------------------|---|--|
| Weightage Distributio n | CA:25%; CE:25%; ESE:50% | |
| Text book/s* | Gupta. S.C. & Kapoor, V.K. (2002): Fundamentals of Mathematical Statistics, Sultan Chand & Sons. Westfall, P., & Henning, K. S. (2013): Understanding Advanced Statistical Methods. CRC Press. | |
| Other Reference s | Croxton, Fredrick E., Cowden, Dudley J. and Klein, S. (1973): Applied General Statistics, 3rd Edition. Prentice Hall. Mukhopadhyay P. (1999): Applied Statistics, Books and Allied Pvt. Ltd. | |

| РО | PO | РО | PO | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA359.1 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| BDA359.2 | 1 | 2 | 3 | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| BDA359.3 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| BDA359.4 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| BDA359.5 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| BDA359.6 | 1 | 2 | 2 | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | 1.0 | 1.0 | 1.0 | 3.0 | 1.0 | 1.0 | 3.0 | 1.0 | 2.0 | 3.0 |



| Sch | ool: SSBSR | Batch: 2024-28 | | | | | | | |
|---------------------|-------------------------------|--|----------------------------|--|--|--|--|--|--|
| | | Academic Year: 2026-27 | | | | | | | |
| (Ho | | Q | | | | | | | |
| | nch: Data Science | Semester: V | | | | | | | |
| & A 1 | nalytics Course Code | BDA363 | | | | | | | |
| - | | | | | | | | | |
| 2 | | Experimental Design Lab | | | | | | | |
| 3 | Credits | | | | | | | | |
| 4 | Contact Hours (L-T-P) | 0-0-2 | | | | | | | |
| | Course Status | DSE | | | | | | | |
| 5 | Course Objectiv e | e course objective is to learn how to plan, design and c periments efficiently and effectively, and analyze the resulting of tain objective conclusions. | | | | | | | |
| 6 | Course | After the completion of this course, the student will be able | | | | | | | |
| | | CO1: Build knowledge of basic principles of design of experiment. CO2: Make use of the concept to various simple types of experimental designs. CO3: Make use of the concept to f complex types of experimental designs. | | | | | | | |
| 7 | Course | CO4: Evaluate the factorial experiment, confounding and s design. CO5: Apply concept of missing-plot techniques, cross-over transformation of data and response question. CO6: How to <i>design</i> and conduct <i>experiments</i> , and how to a properly to answer various <i>research</i> questions The course objective is to learn how to plan, design a | design, and nalyze them | | | | | | |
| | Descriptio | experiments efficiently and effectively, and analyze the resul | | | | | | | |
| | n | obtain objective conclusions. | | | | | | | |
| 8 | Outline syllabus | 8 | CO Mapping | | | | | | |
| | Unit 1 | | CO1 | | | | | | |
| | A, B, C | Problem based on uniformity trial data analysis, formation of plots and blocks. | CO1 | | | | | | |
| | Unit 2 | | | | | | | | |
| | A, B, C | Problem based on Fair field Smith Law, Analysis of data obtained from CRD, RBD, LSD | CO2 | | | | | | |
| | Unit 3 | | | | | | | | |
| | A, B, C | Problem based on analysis of factorial experiments without and with confounding. | CO3 | | | | | | |
| | Unit 4 | | | | | | | | |
| | A, B, C | Problem based on Analysis of Covariance | CO4, CO5 | | | | | | |
| | Unit 5 | | | | | | | | |
| | A, B, C | Analysis with missing data, Split plot and strip plot designs. | CO6 | | | | | | |
| | Mode of examinatio n | Practical+Viva | | | | | | | |
| | Weightage Distributio n | CA:25%; CE:25%; ESE:50% | | | | | | | |



| Те | ext book/s* | 1. Gupta. S.C. & Kapoor, V.K.: Fundamentals of |
|----|-------------|---|
| | | Mathematical Statistics, Sultan Chand & Sons Pvt. Ltd. |
| | | New Delhi. |
| | | 2. Westfall, P., & Henning, K. S.: Understanding Advanced Statistical Methods. CRC Press. |
| Ot | ther | 1.Cochran, Wigand Cox, G.M. Experimental Designs. John |
| Re | eference | Wiley and Sons. |
| s | | 2.Das, M.N. and Giri, Design and Analysis of Experiments. |
| | | New Age International. |

| РО | PO | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|-----|-----|-----|----|----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA363.1 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | | 1 | 2 | |
| BDA363.2 | 1 | 2 | 3 | 2 | | 1 | 1 | 3 | 1 | | | 1 | 2 | |
| BDA363.3 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | | 1 | 2 | |
| BDA363.4 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | | 1 | 2 | |
| BDA363.5 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | | 1 | 2 | |
| BDA363.6 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | | 1 | 2 | |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | | 1.0 | 2.0 | |



| | ool: SSBSR | Batch: 2024-28 | | | | | | | |
|-------------------|--|---|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2026-27 | | | | | | | |
| <u>(Ho</u> Bra | nch: Data Science | & Semester: VI | | | | | | | |
| | lytics | | | | | | | | |
| 1 | Course Code | CMS331 | | | | | | | |
| 2 | Course Title | Numerical Methods | | | | | | | |
| 3 | Credits | 4 | | | | | | | |
| 4 | Contact Hours (L-T-P) | 4-0-0 | | | | | | | |
| | Course Status | Compulsory | | | | | | | |
| 5 | Course Objective | 1. To provide the student with numerical methods of solving the non equations, interpolation, differentiation, and integration. | | | | | | | |
| 6 | Course | 2. To improve the student's skills in numerical methods by using the student will be able to: | he MATLAB. | | | | | | |
| | Outcomes | CO1:Solve a linear system of equations using an appropriation develop the algorithm in MATLAB. (K1,K3,K5,K6) CO2: Solve the algebraic or transcendental equations using num and develop the algorithm in MATLAB. (K1,K3,K5,K6) CO3: Discuss the finite difference methods to analyse the function CO4: Explain the divided difference and evaluate the function. (K2) CO5:Describe the numerical differentiation and evaluate the difference K2, K5) CO6: Calculate a definite integral using an appropriation method at the function of the system of the system. | nerical methods s (K2,K4) 2, K4, K5) rentiation. (K1, | | | | | | |
| | | algorithm in MATLAB. (K1,K3,K5,K6) | | | | | | | |
| 7 | Course Description | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numeric and skills to implement algorithms to solve mathematical MATLAB | cal algorithms | | | | | | |
| 7 8 | | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numeric and skills to implement algorithms to solve mathematical MATLAB. | cal algorithms problems in | | | | | | |
| | Description | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numeric and skills to implement algorithms to solve mathematical MATLAB. | cal algorithms problems in | | | | | | |
| | Description Outline syllabus | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numeric and skills to implement algorithms to solve mathematical MATLAB | cal algorithms problems in | | | | | | |
| | Description Outline syllabus Unit 1 | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numeric and skills to implement algorithms to solve mathematical MATLAB Solution of system of linear equations: | cal algorithms problems in CO Mapping | | | | | | |
| | Description Outline syllabus Unit 1 A B | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numeric and skills to implement algorithms to solve mathematical MATLAB Solution of system of linear equations: Direct methods: Cramer's rule, Matrix inverse method Gauss elimination and Gauss-Jordan method | cal algorithms problems in CO Mapping CO1 CO1 | | | | | | |
| | Description Outline syllabus Unit 1 A B C | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numeric and skills to implement algorithms to solve mathematical MATLAB Solution of system of linear equations: Direct methods: Cramer's rule, Matrix inverse method Gauss elimination and Gauss-Jordan method Iterative methods: Jacobi's method, Gauss-Seidal method | cal algorithms problems in CO Mapping CO1 | | | | | | |
| | Description Outline syllabus Unit 1 A B | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numerical and skills to implement algorithms to solve mathematical MATLAB Solution of system of linear equations: Direct methods: Cramer's rule, Matrix inverse method Gauss elimination and Gauss-Jordan method Iterative methods: Jacobi's method, Gauss-Seidal method System of Transcendental equations: Initial approximation of the roots, Bisection method, Method or | cal algorithms problems in CO Mapping CO1 CO1 CO1 | | | | | | |
| | Description Outline syllabus Unit 1 A B C Unit 2 | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numeric and skills to implement algorithms to solve mathematical MATLAB Solution of system of linear equations: Direct methods: Cramer's rule, Matrix inverse method Gauss elimination and Gauss-Jordan method Iterative methods: Jacobi's method, Gauss-Seidal method System of Transcendental equations: | cal algorithms problems in CO Mapping CO1 CO1 CO1 | | | | | | |
| | Description Outline syllabus Unit 1 A B C Unit 2 A | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numerica and skills to implement algorithms to solve mathematical MATLAB Solution of system of linear equations: Direct methods: Cramer's rule, Matrix inverse method Gauss elimination and Gauss-Jordan method Iterative methods: Jacobi's method, Gauss-Seidal method System of Transcendental equations: Initial approximation of the roots, Bisection method, Method of false position | al algorithms problems in CO Mapping CO1 CO1 CO1 CO1 fCO2 | | | | | | |
| | Description Outline syllabus Unit 1 A B C Unit 2 A B | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numeric and skills to implement algorithms to solve mathematical MATLAB Solution of system of linear equations: Direct methods: Cramer's rule, Matrix inverse method Gauss elimination and Gauss-Jordan method Iterative methods: Jacobi's method, Gauss-Seidal method System of Transcendental equations: Initial approximation of the roots, Bisection method, Method of false position Secant method, iteration method, Newton-Raphson method and its convergence. Finite differences and Interpolation | al algorithms problems in CO Mapping CO1 CO1 CO1 fCO2 CO2 CO2 | | | | | | |
| | Description Outline syllabus Unit 1 A B C Unit 2 A B C C | This course is an introduction to the numerical analysis. The print of the course is to develop the basic understanding of numerical and skills to implement algorithms to solve mathematical MATLAB Solution of system of linear equations: Direct methods: Cramer's rule, Matrix inverse method Gauss elimination and Gauss-Jordan method Iterative methods: Jacobi's method, Gauss-Seidal method System of Transcendental equations: Initial approximation of the roots, Bisection method, Method of false position Secant method, iteration method, Newton-Raphson method and its convergence. Finite differences and Interpolation Finite difference operators, their properties and their interrelations, finite difference tables. | al algorithms problems in CO Mapping CO1 CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO2 | | | | | | |
| | Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 2 C Unit 3 | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numeric and skills to implement algorithms to solve mathematical MATLAB Solution of system of linear equations: Direct methods: Cramer's rule, Matrix inverse method Gauss elimination and Gauss-Jordan method Iterative methods: Jacobi's method, Gauss-Seidal method System of Transcendental equations: Initial approximation of the roots, Bisection method, Method of false position Secant method, iteration method, Newton-Raphson method and its convergence. Finite differences and Interpolation | al algorithms problems in CO Mapping CO1 CO1 CO1 fCO2 CO2 CO2 | | | | | | |
| | Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 2 A B C Unit 3 A | This course is an introduction to the numerical analysis. The print of the course is to develop the basic understanding of numerical and skills to implement algorithms to solve mathematical MATLAB Solution of system of linear equations: Direct methods: Cramer's rule, Matrix inverse method Gauss elimination and Gauss-Jordan method Iterative methods: Jacobi's method, Gauss-Seidal method System of Transcendental equations: Initial approximation of the roots, Bisection method, Method of false position Secant method, iteration method, Newton-Raphson method and its convergence. Finite differences and Interpolation Finite difference operators, their properties and their interrelations, finite difference tables. Newton's forward and Newton's backward interpolation formula Central difference formulae including Stirling's formula, Bessel' formula. | al algorithms problems in CO Mapping CO1 CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO2 CO3 CO3 | | | | | | |
| | Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 3 A B | This course is an introduction to the numerical analysis. The prin of the course is to develop the basic understanding of numeric and skills to implement algorithms to solve mathematical MATLAB S Solution of system of linear equations: Direct methods: Cramer's rule, Matrix inverse method Gauss elimination and Gauss-Jordan method Iterative methods: Jacobi's method, Gauss-Seidal method System of Transcendental equations: Initial approximation of the roots, Bisection method, Method of false position Secant method, iteration method, Newton-Raphson method and its convergence. Finite difference operators, their properties and their interrelations, finite difference tables. Newton's forward and Newton's backward interpolation formula Central difference formulae including Stirling's formula, Bessel' | al algorithms problems in CO Mapping CO1 CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO2 CO3 CO3 | | | | | | |
| | Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 3 A B C C Unit 3 C C | This course is an introduction to the numerical analysis. The print of the course is to develop the basic understanding of numerical and skills to implement algorithms to solve mathematical MATLAB Solution of system of linear equations: Direct methods: Cramer's rule, Matrix inverse method Gauss elimination and Gauss-Jordan method Iterative methods: Jacobi's method, Gauss-Seidal method System of Transcendental equations: Initial approximation of the roots, Bisection method, Method of false position Secant method, iteration method, Newton-Raphson method and its convergence. Finite differences and Interpolation Finite difference operators, their properties and their interrelations, finite difference tables. Newton's forward and Newton's backward interpolation formula Central difference formulae including Stirling's formula, Bessel' formula. | al algorithms problems in CO Mapping CO1 CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO2 CO3 CO3 | | | | | | |



| С | Lagrange's interpolation formula. | CO4 |
|---------------------------|--|-----|
| Unit 5 | Numerical differentiation and integration | |
| А | Differentiation using Newton's forward and backward formula | CO5 |
| В | Newton-Cotes Quadrature formula -derivations & comparison of Trapezoidal rule | CO6 |
| С | Simpson's 1/3 and 3/8 rules. | CO6 |
| Mode of examination | Theory | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | 1.EndreSuli, David F. Mayers: An Introduction to Numerical Analysis. Cambridge University Press. 2.Gupta, R.S.: Elements of Numerical Analysis Macmillan India Ltd. | |
| Other References | 1. Grewal, B.S. Numerical methods in Engineering & Science Khanna Publishers. | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | PO | РО | PO | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| CMS331.1 | 3 | 3 | 2 | 2 | 2 | 1 | | | | | 2 | | | |
| CMS331.2 | 2 | 2 | 2 | 2 | 2 | 1 | | | | | 2 | | | |
| CMS331.3 | 3 | 3 | 3 | 3 | 2 | 1 | | | | | 2 | | | |
| CMS331.4 | 2 | 2 | 2 | 3 | 2 | 1 | | | | | 2 | | | |
| CMS331.5 | 2 | 3 | 3 | 3 | 2 | 1 | | | | | 2 | | | |
| CMS331.6 | 3 | 2 | 3 | 3 | 2 | 1 | | | | | 2 | | | |
| Average | 2.5 | 2.5 | 2.5 | 2.6 | 2.0 | 1.0 | | | | | 2.0 | | | |



| Scho | ool: SSBSR | Batch: 2024-28 | | | | | | |
|------|--|---|---|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2026-27 | | | | | | |
| (Hor | | | | | | | | |
| | nch: Data Science | Semester: VI | | | | | | |
| | nalytics | DD 4 222 | | | | | | |
| 1 | Course Code | BDA322 | | | | | | |
| 2 | Course Title | Statistical Simulation | | | | | | |
| 3 | Credits | 4 | | | | | | |
| 4 | Contact Hours (L-T-P) | 4-0-0 | | | | | | |
| | Course Status | CC | | | | | | |
| 5 | Course Objective | The learning objectives include: Concept of simulation and simulati Generation of Pseudo random number generators as well as fri statistical distributions, Monte-Carlo simulation technique and a simulation techniques. | om standard | | | | | |
| 6 | Course Outcomes | CO1: Recognize the concepts of probability and statistics that are relevant to modeling and simulation (K2, K3). CO2: How to generate random numbers by the different methods (K2, K3). CO3: Design and implement Bootstrapping; jackknife resampling(K3, K4). CO4: Be able to evaluate and interpret the Markov-Chain Monte Carlo (MCMC) simulations (K3, K4). CO5: Hands-on experience in using simulation software packages/structured programming languages (K3, K4, K5) CO6: How simulation may be used to understand the behavior of real world systems by utilizing mathematical models with an emphasis on simulation (K4, | | | | | | |
| | | | | | | | | |
| 7 | Course Description | systems by utilizing mathematical models with an emphasis on sin | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling | | | | | |
| 7 | | systems by utilizing mathematical models with an emphasis on sin K6). The course topics will include a review of concepts from probability that are relevant to modeling and simulation, algorithms for ran sampling, modeling and analysis of basic queueing systems, variatechniques, statistical-validation techniques, Independent Monte Carl Markov-Chain Monte Carlo (MCMC) simulations, and discrete-ev and simulation. | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling CO | | | | | |
| | Description Outline syllabus | systems by utilizing mathematical models with an emphasis on sin K6). The course topics will include a review of concepts from probability that are relevant to modeling and simulation, algorithms for ran sampling, modeling and analysis of basic queueing systems, variatechniques, statistical-validation techniques, Independent Monte Carl Markov-Chain Monte Carlo (MCMC) simulations, and discrete-ev and simulation. | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling | | | | | |
| | Description Outline syllabus Unit 1 | systems by utilizing mathematical models with an emphasis on sin K6). The course topics will include a review of concepts from probability that are relevant to modeling and simulation, algorithms for ran sampling, modeling and analysis of basic queueing systems, variatechniques, statistical-validation techniques, Independent Monte Carlo Markov-Chain Monte Carlo (MCMC) simulations, and discrete-ev and simulation. | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling CO Mapping | | | | | |
| | Description Outline syllabus | systems by utilizing mathematical models with an emphasis on sin K6). The course topics will include a review of concepts from probability that are relevant to modeling and simulation, algorithms for ran sampling, modeling and analysis of basic queueing systems, variatechniques, statistical-validation techniques, Independent Monte Carl Markov-Chain Monte Carlo (MCMC) simulations, and discrete-ev and simulation. | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling CO | | | | | |
| | Description Outline syllabus Unit 1 A | systems by utilizing mathematical models with an emphasis on sin K6). The course topics will include a review of concepts from probability that are relevant to modeling and simulation, algorithms for ran sampling, modeling and analysis of basic queueing systems, varia techniques, statistical-validation techniques, Independent Monte Carl Markov-Chain Monte Carlo (MCMC) simulations, and discrete-ev and simulation. Review of R/Python. Random number generation: Inverse-transform; acceptance- | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling CO Mapping CO1 | | | | | |
| | Description Outline syllabus Unit 1 A B | systems by utilizing mathematical models with an emphasis on sin K6). The course topics will include a review of concepts from probability that are relevant to modeling and simulation, algorithms for ran sampling, modeling and analysis of basic queueing systems, varia techniques, statistical-validation techniques, Independent Monte Carl Markov-Chain Monte Carlo (MCMC) simulations, and discrete-ev and simulation. Review of R/Python. Random number generation: Inverse-transform; acceptance- rejection; transformations. Statistic simulations: generating random variables, simulating | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling CO Mapping CO1 CO1 | | | | | |
| | Description Description Outline syllabus Unit 1 A B C | systems by utilizing mathematical models with an emphasis on sin K6). The course topics will include a review of concepts from probability that are relevant to modeling and simulation, algorithms for ran sampling, modeling and analysis of basic queueing systems, varia techniques, statistical-validation techniques, Independent Monte Carl Markov-Chain Monte Carlo (MCMC) simulations, and discrete-ev and simulation. Review of R/Python. Random number generation: Inverse-transform; acceptance- rejection; transformations. Statistic simulations: generating random variables, simulating | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling CO Mapping CO1 CO1 | | | | | |
| | Description Outline syllabus Unit 1 A B C Unit 2 | systems by utilizing mathematical models with an emphasis on sin K6). The course topics will include a review of concepts from probability that are relevant to modeling and simulation, algorithms for ran sampling, modeling and analysis of basic queueing systems, variate techniques, statistical-validation techniques, Independent Monte Carl Markov-Chain Monte Carlo (MCMC) simulations, and discrete-ev and simulation. Review of R/Python. Random number generation: Inverse-transform; acceptance-rejection; transformations. Statistic simulations: generating random variables, simulating normal, gamma and beta random variables. | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling CO Mapping CO1 CO1 CO1 | | | | | |
| | Description Outline syllabus Unit 1 A B C Unit 2 A | systems by utilizing mathematical models with an emphasis on sin K6). The course topics will include a review of concepts from probability that are relevant to modeling and simulation, algorithms for ran sampling, modeling and analysis of basic queueing systems, variat techniques, statistical-validation techniques, Independent Monte Carl Markov-Chain Monte Carlo (MCMC) simulations, and discrete-ev and simulation. Review of R/Python. Random number generation: Inverse-transform; acceptance-rejection; transformations. Statistic simulations: generating random variables, simulating normal, gamma and beta random variables. Simulating multivariate distributions, MCMC methods. Gibbs sampler, simulating random fields, simulating stochastic | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling CO1 CO1 CO1 CO1 | | | | | |
| | Description Description Outline syllabus Unit 1 A B C Unit 2 A B | systems by utilizing mathematical models with an emphasis on sin K6). The course topics will include a review of concepts from probability that are relevant to modeling and simulation, algorithms for ran sampling, modeling and analysis of basic queueing systems, variatechniques, statistical-validation techniques, Independent Monte Carl Markov-Chain Monte Carlo (MCMC) simulations, and discrete-evand simulation. Review of R/Python. Random number generation: Inverse-transform; acceptance-rejection; transformations. Statistic simulations: generating random variables, simulating normal, gamma and beta random variables. Simulating multivariate distributions, MCMC methods. Gibbs sampler, simulating random fields, simulating stochastic process. Variance reduction technique: importance sampling for integration, | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling CO1 CO1 CO1 CO1 CO2 CO2 | | | | | |
| | Description Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 2 A C C C C C | systems by utilizing mathematical models with an emphasis on sin K6). The course topics will include a review of concepts from probability that are relevant to modeling and simulation, algorithms for ran sampling, modeling and analysis of basic queueing systems, variatechniques, statistical-validation techniques, Independent Monte Carl Markov-Chain Monte Carlo (MCMC) simulations, and discrete-evand simulation. Review of R/Python. Random number generation: Inverse-transform; acceptance-rejection; transformations. Statistic simulations: generating random variables, simulating normal, gamma and beta random variables. Simulating multivariate distributions, MCMC methods. Gibbs sampler, simulating random fields, simulating stochastic process. Variance reduction technique: importance sampling for integration, | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling CO1 CO1 CO1 CO1 CO2 CO2 | | | | | |
| | Description Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 2 A B C Unit 3 | systems by utilizing mathematical models with an emphasis on sin K6). The course topics will include a review of concepts from probability that are relevant to modeling and simulation, algorithms for ran sampling, modeling and analysis of basic queueing systems, varia techniques, statistical-validation techniques, Independent Monte Car Markov-Chain Monte Carlo (MCMC) simulations, and discrete-ev and simulation. Review of R/Python. Random number generation: Inverse-transform; acceptance- rejection; transformations. Statistic simulations: generating random variables, simulating normal, gamma and beta random variables. Simulating multivariate distributions, MCMC methods. Gibbs sampler, simulating random fields, simulating stochastic process. Variance reduction technique: importance sampling for integration, control variates and antithetic variables. | nulation (K4, and statistics dom-variable nce-reduction lo (IMC) and ent modeling CO1 CO1 CO1 CO1 CO2 CO2 CO2 | | | | | |



| Unit 4 | | |
|---------------------|--|----------|
| А | Simulating a non-homogeneous Poisson process. | CO4 |
| В | Optimization using Monte Carlo methods simulated annealing for optimization. | CO4 |
| С | Solving differential equations by Monte Carlo methods. | CO4 |
| Unit 5 | | |
| А | Univariate density estimation; kernel smoothing multivariate density estimation | CO5, CO6 |
| В | Root finding; more on numerical integration; numerical maximization/minimization; constrained and unconstrained optimization. | CO5, CO6 |
| С | EM (Expectation-Maximization) algorithm; simplex algorithm. | CO5, CO6 |
| Mode of examina | 5 | |
| Weighta Distribu | \sim CA:25%: ESE:75% | |
| Text boo | bk/s* 1.Fishman, G.S. Monte Carlo: Concepts, Algorithms and Applications. 2.Rubinstein, R.Y.: Simulation and the Monte Carlo Method. | |
| Other | 1.Ross, S. M.: Simulation, Third Edition, Academic Press. | |
| Reference | ces 2.Efron, B. and Tibshirani. R.J.: An introduction to the Bootstrap. | |

| РО | PO | PO | РО | РО | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA322.1 | 3 | 3 | 2 | 2 | 2 | 1 | | | | | 2 | 2 | | |
| BDA322.2 | 2 | 3 | 3 | 2 | 2 | 1 | | | | | 2 | 2 | | |
| BDA322.3 | 2 | 2 | 2 | 3 | 2 | 1 | | | | | 2 | 2 | | |
| BDA322.4 | 2 | 3 | 2 | 2 | 2 | 1 | | | | | 2 | 2 | | |
| BDA322.5 | 3 | 3 | 2 | 2 | 2 | 1 | | | | | 2 | 2 | | |
| BDA322.6 | 3 | 3 | 2 | 3 | 2 | 1 | | | | | 2 | 2 | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | 2.0 | 1.0 | | | | | 2.0 | 2.0 | | |



| | ool: SSBSR | Batch: 2024-28 | | | | | | | |
|------|-------------------------------|---|----------------------------------|--|--|--|--|--|--|
| (Hoi | | Academic Year: 2026-27 | | | | | | | |
| | nch: Data Science nalytics | | | | | | | | |
| 1 | Course Code | BDA323 | | | | | | | |
| 2 | Course Title | Multivariate Data Analysis | | | | | | | |
| 3 | Credits | 3 | | | | | | | |
| 4 | Contact Hours (L-T-P) | 3-0-0 | | | | | | | |
| | Course Status | CC | | | | | | | |
| 5 | Course Objective | Familiarise students with the multivariate normal distribution, estimation of the mean vector and the covariance matrix, the distributions and uses of sample correlation coefficients, classification of observations, the distribution of the sample covariance matrix, and the sample generalized variance. | | | | | | | |
| 6 | Course Outcomes | CO1: Demonstrate knowledge and understanding of the multivariate normal distribution. (K2, K3) CO2: Demonstrate knowledge and understanding of the concept of estimation of the mean vector and the covariance matrix. (K2, K3) CO3: Demonstrate advanced understanding of the concepts of dimension reduction technique. (K2, K3) CO4: Describe the concepts of how to use and apply dependence techniques in multivariate data analysis. (K2, K3) CO5: Describe the concepts of analysis of variance and covariance in multivariate data analysis. (K3, K4, K5) CO6: Apply the statistical tool and software in multivariate data analysis. (K2, K2) | | | | | | | |
| 7 | Course Description | K6) This module aims to provide an understanding of the multiv distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, cla observations, the distribution of the sample covariance matrix, and generalized variance. | e matrix, the assification of | | | | | | |
| 8 | | | | | | | | | |
| | Unit 1 A | A brief review of Univariate and Bivariate distribution with their | CO1 | | | | | | |
| | В | properties. Basic Multivariate Distribution: mean, variance, Covariance, correlation, and the linear combination of variables. | CO1 | | | | | | |
| | С | The multivariate normal distribution, Mean Vectors, and Covariance Matrices. | CO1 | | | | | | |
| | Unit 2 | | | | | | | | |
| | А | Multivariate normal distribution; maximum likelihood estimation, Wishart's distribution | CO2 | | | | | | |
| | В | Hotelling's T2 and hypothesis testing for multivariate normal data. Inference from a single sample, Inference from two dependent samples Inference from two independent samples. | CO2 | | | | | | |
| | С | Simple, Multiple, Partial, and Canonical correlations with their properties. | CO2 | | | | | | |
| | Unit 3 | | | | | | | | |
| | А | Principal Components Analysis and derivation of principal components; PCA structural model; PCA on normal populations; bi-plots. | CO3 | | | | | | |
| | В | Factor Analysis, Factor extraction Factor rotation, Factor scores Validation of factor analysis, Higher order factor analysis Q-type factor analysis | CO3, CO4 | | | | | | |



| С | Cluster Analysis, Types of clustering, Correlation, and distance, Partitioning methods, hierarchical clustering, K-means clustering, and their interpretation. | CO4 |
|---------------------------|---|-----|
| Unit 4 | | |
| А | Simple, Multiple, and Multivariate regression with their properties. | CO5 |
| В | Binary and multidimensional Logistic regression. | CO5 |
| С | Linear discriminant function analysis. Estimating linear discriminant functions and their properties. | CO5 |
| Unit 5 | | |
| А | Analysis of variance and covariance. | CO6 |
| В | Multivariate analysis of variance and Covariance. | CO6 |
| С | Concepts of correspondence analysis, chi-square distance and inertia, multiple correspondence analysis. | CO6 |
| Mode of examination | Theory | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | I.Johnson, R.A. and Wichern, D.W.: Applied Multivariate Statistical Analysis, Sixth Edition, Pearson Education India. 2.Hardle, W.K. and Hlavka, Z. Multivariate Statistics, Springer. | |
| Other References | Anderson, T.W.: An Introduction to Multivariate Statistical Analysis, Third Edition, Wiley. Härdle, W.K. and Simar, L. : Applied Multivariate Statistical Analysis, Springer. | |

| РО | РО | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA323.1 | 3 | 3 | 2 | 2 | 2 | 1 | | | | | 2 | 2 | | |
| BDA323.2 | 2 | 3 | 3 | 2 | 2 | 1 | | | | | 2 | 2 | | |
| BDA323.3 | 2 | 2 | 2 | 3 | 2 | 1 | | | | | 2 | 2 | | |
| BDA323.4 | 2 | 3 | 2 | 2 | 2 | 1 | | | | | 2 | 2 | | |
| BDA323.5 | 3 | 3 | 2 | 2 | 2 | 1 | | | | | 2 | 2 | | |
| BDA323.6 | 3 | 3 | 2 | 3 | 2 | 1 | | | | | 2 | 2 | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | 2.0 | 1.0 | | | | | 2.0 | 2.0 | | |



| Sch | ool: SSBSR | Batch: 2024-28 | |
|-----|-------------------------------|---|------------------|
| (Ho | | Academic Year: 2026-27 | |
| & A | nch: Data Science nalytics | | |
| 1 | Course Code | RBL004 | |
| 2 | Course Title | Research Based Learning-4 | |
| 3 | Credits | 1 | |
| 4 | Contact Hours | 0-0-2 | |
| | (L-T-P) | | |
| | Course Status | Project (Audit-Qualifying) | |
| 5 | Course | 1. Deep knowledge of a specific area of specialization. | |
| | Objective | 2. Develop communication skills, especially in project w | riting and oral |
| (| | presentation. Develop some time management skills. | 1 1 . |
| 6 | Course Outcomes | CO1: Explain the concept of research within the subject, as regard question, collecting and analyzing background material, and pre | |
| | | questions and conclusions. (K2, K4) | - |
| | | CO2: Construct and develop a deeper interest in mathematics | and a taste for |
| | | research. (K5, K6) | |
| | | CO3: Select and recommend activities that support their profession | onal goals. (K4, |
| | | | |
| | | CO4: Develop effective project organizational skills. (K5) | 75) |
| | | CO5: Analyse the problem and summarize research findings. (K4,F CO6: Use research findings to develop education theory and practic | |
| 7 | Course | Maintain a core of mathematical and technical knowledge that is ad | |
| , | Description | changing technologies and provides a solid foundation for future le | |
| 8 | | 1 | |
| | Unit 1 | Introduction | CO1 |
| | Unit 2 | Case study | CO1,CO2 |
| | | | |
| | Unit 3 | Conceptual | CO3,CO4 |
| | | | |
| | Unit 4 | Development | CO4, CO5 |
| | Unit 5 | Finalisation | CO5, CO6 |
| | Mode of | | |
| | examination | | |
| | Weightage | | 1 |
| | Distribution | | |
| | Text book/s* | | |
| | Other | | |
| | References | | |



| РО | РО | PO | РО | PO | PO | PO | PO | РО | РО | РО | РО | PSO | PSO | PSO |
|----------|----|----|----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| RBL004.1 | | | | 2 | 3 | 3 | 3 | 3 | 3 | 3 | 3 | 3 | 1 | 1 |
| RBL004.2 | | | | 2 | 3 | 3 | 3 | 3 | 3 | 3 | 3 | 3 | 1 | 1 |
| RBL004.3 | | | | 2 | 3 | 3 | 3 | 3 | 3 | 3 | 3 | 3 | 1 | 1 |
| RBL004.4 | | | | 2 | 3 | 3 | 3 | 3 | 3 | 3 | 3 | 3 | 1 | 1 |
| RBL004.5 | | | | 2 | 3 | 3 | 3 | 3 | 3 | 3 | 3 | 3 | 1 | 1 |
| RBL004.6 | | | | 2 | 3 | 3 | 3 | 3 | 3 | 3 | 3 | 3 | 1 | 1 |
| Average | | | | 2.0 | 3.0 | 3.0 | 3.0 | 3.0 | 3.0 | 3.0 | 3.0 | 3.0 | 1.0 | 1.0 |



| | ool: SSBSR | Batch: 2024-28 | |
|------|--|---|--|
| | gramme: B.Sc. | Academic Year: 2026-27 | |
| (Hoi | | | |
| | nch: Data Science | Semester: VI | |
| | nalytics Course Code | DD 4 225 | |
| 1 | | BDA325 | |
| 2 | Course Title | Deep Learning | |
| 3 | Credits | 3 | |
| 4 | Contact Hours | 3-0-0 | |
| | (L-T-P) | 5-0-0 | |
| | Course Status | CC | |
| 5 | Course | The objective of this course is to provide advance knowledge of D | Deep learning |
| | Objective | techniques and also apply Deep learning Techniques to various eng | gineering and |
| | - | social applications. | |
| 6 | Course | CO1: Ability to identify the deep learning techniques (K2, K3). | |
| | Outcomes | CO2: Ability to select and implement Machine learning and d | eep learning |
| | | (K2,K3,K4) | |
| | | CO3: Ability to Train machine and solve problems associated with b | atch learning |
| | | and online learning (K2, K3, K4). | |
| | | CO4: Ability to recognize and implement various ways of selecting su | uitable model |
| | | parameters for different machine learning techniques(K3, K4,K5). | |
| | | CO5: Ability to integrate deep learning libraries and mathematical a | and statistical |
| | | tools(K4, K5). | |
| | | | |
| | | CO6: Ability to apply Deep learning Techniques to various engineering | ng and social |
| | | CO6: Ability to apply Deep learning Techniques to various engineerin applications(K4, K6). | ng and social |
| | | applications(K4, K6). | |
| 7 | Course | applications(K4, K6). This course mainly focused on Regression and Neural network ba | sed Machine |
| 7 | Course Description | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va | sed Machine |
| | | applications(K4, K6). This course mainly focused on Regression and Neural network ba | sed Machine |
| 7 | Description | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va | sed Machine |
| | Description Unit 1 | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. | sed Machine rious recent |
| | Description | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. | sed Machine rious recent CO1 |
| | Description Unit 1 | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, | sed Machine rious recent |
| | Description Unit 1 A B | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. | sed Machine rious recent <u>CO1</u> CO1 |
| | Description Unit 1 A | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, | sed Machine rious recent CO1 |
| | Description Unit 1 A B | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. | sed Machine rious recent <u>CO1</u> CO1 |
| | Description Unit 1 A B C | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. | sed Machine rious recent <u>CO1</u> CO1 |
| | Description Unit 1 A B C Unit 2 | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. | sed Machine rious recent CO1 CO1 CO1 |
| | Description Unit 1 A B C Unit 2 | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov | sed Machine rious recent CO1 CO1 CO1 |
| | Description Unit 1 A B C Unit 2 A | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov Accelerated GD, Stochastic GD. Principal Component Analysis and its interpretations, Singular Value Decomposition. | sed Machine rious recent CO1 CO1 CO2 CO2 |
| | Description Unit 1 A B C Unit 2 A | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov Accelerated GD, Stochastic GD. Principal Component Analysis and its interpretations, Singular | sed Machine rious recent CO1 CO1 CO1 CO2 |
| | Description Unit 1 A B C Unit 2 A B C C | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov Accelerated GD, Stochastic GD. Principal Component Analysis and its interpretations, Singular Value Decomposition. | sed Machine rious recent CO1 CO1 CO2 CO2 |
| | Description Unit 1 A B C Unit 2 A B B | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov Accelerated GD, Stochastic GD. Principal Component Analysis and its interpretations, Singular Value Decomposition. Auto encoders and relation to PCA, Regularization in auto encoders, | sed Machine rious recent CO1 CO1 CO2 CO2 |
| | Description Unit 1 A B C Unit 2 A B C C | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov Accelerated GD, Stochastic GD. Principal Component Analysis and its interpretations, Singular Value Decomposition. Auto encoders and relation to PCA, Regularization in auto encoders, | sed Machine rious recent CO1 CO1 CO2 CO2 |
| | Description Unit 1 A B C Unit 2 A B C Unit 2 Unit 3 | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov Accelerated GD, Stochastic GD. Principal Component Analysis and its interpretations, Singular Value Decomposition. Auto encoders and relation to PCA, Regularization in auto encoders, Denoising auto encoders, Sparse auto encoders. Regularization: Bias Variance Tradeoff, L2 regularization, Early stopping, Data set augmentation. | sed Machine rious recent CO1 CO1 CO2 CO2 CO2 |
| | Description Unit 1 A B C Unit 2 A B C Unit 2 Unit 3 | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov Accelerated GD, Stochastic GD. Principal Component Analysis and its interpretations, Singular Value Decomposition. Auto encoders and relation to PCA, Regularization in auto encoders, Denoising auto encoders, Sparse auto encoders. Regularization: Bias Variance Tradeoff, L2 regularization, Early | sed Machine rious recent CO1 CO1 CO2 CO2 CO2 |
| | Description Unit 1 A B C Unit 2 A B C Unit 2 A B C Unit 3 A | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov Accelerated GD, Stochastic GD. Principal Component Analysis and its interpretations, Singular Value Decomposition. Auto encoders and relation to PCA, Regularization in auto encoders, Denoising auto encoders, Sparse auto encoders. Regularization: Bias Variance Tradeoff, L2 regularization, Early stopping, Data set augmentation. | sed Machine rious recent CO1 CO1 CO2 CO2 CO2 CO2 CO2 |
| | Description Unit 1 A B C Unit 2 A B C Unit 2 A B C Unit 3 A | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov Accelerated GD, Stochastic GD. Principal Component Analysis and its interpretations, Singular Value Decomposition. Auto encoders and relation to PCA, Regularization in auto encoders, Denoising auto encoders, Sparse auto encoders. Regularization: Bias Variance Tradeoff, L2 regularization, Early stopping, Data set augmentation. Greedy Layer wise Pre-training, Better activation functions, Better | sed Machine rious recent CO1 CO1 CO2 CO2 CO2 CO2 CO2 |
| | Description Unit 1 A B C Unit 2 A B C Unit 2 A B C Unit 3 A B C C | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov Accelerated GD, Stochastic GD. Principal Component Analysis and its interpretations, Singular Value Decomposition. Auto encoders and relation to PCA, Regularization in auto encoders, Denoising auto encoders, Sparse auto encoders. Regularization: Bias Variance Tradeoff, L2 regularization, Early stopping, Data set augmentation. Greedy Layer wise Pre-training, Better activation functions, Better weight initialization methods, Batch Normalization. | sed Machine rious recent CO1 CO1 CO2 CO2 CO2 CO2 CO2 CO3 CO3 |
| | Description Unit 1 A B C Unit 2 A B C Unit 2 A B C Unit 3 A B C C Unit 4 | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov Accelerated GD, Stochastic GD. Principal Component Analysis and its interpretations, Singular Value Decomposition. Auto encoders and relation to PCA, Regularization in auto encoders, Denoising auto encoders, Sparse auto encoders. Regularization: Bias Variance Tradeoff, L2 regularization, Early stopping, Data set augmentation. Greedy Layer wise Pre-training, Better activation functions, Better weight initialization methods, Batch Normalization. Learning Vectorial Representations Of Words. | sed Machine rious recent CO1 CO1 CO2 CO2 CO2 CO2 CO3 CO3 CO3 |
| | Description Unit 1 A B C Unit 2 A B C Unit 2 A B C Unit 3 A B C C | applications(K4, K6). This course mainly focused on Regression and Neural network ba learning algorithms. This aim to make students aware of va developments in the field of Neural network such as deep learning. History of Deep Learning, McCulloch Pitts Neuron. Multilayer Perceptron's (MLPs), Representation Power of MLPs, Sigmoid Neurons, Gradient Descent. Feed Forward Neural Networks, Back propagation. Gradient Descent (GD), Momentum Based GD, Nesterov Accelerated GD, Stochastic GD. Principal Component Analysis and its interpretations, Singular Value Decomposition. Auto encoders and relation to PCA, Regularization in auto encoders, Denoising auto encoders, Sparse auto encoders. Regularization: Bias Variance Tradeoff, L2 regularization, Early stopping, Data set augmentation. Greedy Layer wise Pre-training, Better activation functions, Better weight initialization methods, Batch Normalization. | sed Machine rious recent CO1 CO1 CO2 CO2 CO2 CO2 CO2 CO3 CO3 |



| В | Recurrent Neural Networks, Back propagation through time (BPTT), Vanishing and Exploding Gradients, Truncated BPTT, GRU, LSTMs. | CO4 |
|---------------------------|---|----------|
| С | Encoder Decoder Models, Attention Mechanism, Attention over images. | CO4 |
| Unit 5 | | |
| А | Advanced Deep architectures: Recurrent Neural networks (RNNs), Generative Adversarial Networks (GANs). | CO5, CO6 |
| В | In-depth discussion of DL examples. | CO5, CO6 |
| С | Advanced topics, Recent papers, Influential papers: Visual Question Answering, Visual Dialog, Novel deep methods (Deep internal learning, Deep image prior). | |
| Mode of examination | Practical Based | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | 1. Mahmoud Hassaballah, Ali Ismail Awad: Deep Learning in Computer Vision, Principles and Applications. | |
| | 2.Dr.P.S. Jagadeesh Kumar, Prof. Thomas Binford, Dr. J. Ruby, J. | |
| | Lepika. Modern Deep Learning and Advanced Computer Vision, A Perspective Approach. | |
| Other References | I.Ian Goodfellow, YoshuaBengio, Aaron Courville, "Deep Learning Adaptive Computation and Machine Learning series", MIT Press. Li Deng and Dong Yu "Deep Learning Methods and Applications", Foundations and Trends in Signal Processing. | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | РО | РО | PO | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|-----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA325.1 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | | |
| BDA325.2 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | | |
| BDA325.3 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | | |
| BDA325.4 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | | |
| BDA325.5 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | | |
| BDA325.6 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | | |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 3.0 | 3.0 | | |



| Sch | ool: SSBSR | Batch: 2024-2 | 28 | | | | | | | |
|----------|-------------------|---|------------------------------|--|------------|--|--|--|--|--|
| Prog | gram: B.Sc. | Academic Y | ear: 2026-27 | | | | | | | |
| <u>`</u> | search) | | | | | | | | | |
| | nch: Data Science | Semester: VI | | | | | | | | |
| | nalytics | | | | | | | | | |
| 1 | Course Code | CCU108 | ~ | | | | | | | |
| 2 | Course Title | Community C | | | | | | | | |
| 3 | Credits | 2 | Course Status: Training | /Survey/Project | | | | | | |
| 4 | (L-T-P) | (0-0-4) | | | | | | | | |
| 5 | Learning | | Contact Hours | 30 | | | | | | |
| | Hours | | Project/Field Work | 20 | | | | | | |
| | | | Assessment | 00 | | | | | | |
| | | | Guided Study | 10 | | | | | | |
| | | | Total hours | 60 | | | | | | |
| 6 | Course | 1. Contr | ibute to the holistic dev | elopment of students by mal | king them | | | | | |
| | Objectives | more | aware of socially and ec | onomically disadvantaged con | mmunities | | | | | |
| | | | eir specific issues | | | | | | | |
| | | | 1 | grand to make them men | offortivo | | | | | |
| | | | | ssrooms, to make them more | | | | | | |
| | | | | gning them to social realitie | es beyond | | | | | |
| | | textbo | ooks | | | | | | | |
| | | 3. Provi | de scope to faculty m | embers to align their teac | hing and | | | | | |
| | | resea | rch goals by giving the | em ample opportunity to | carry out | | | | | |
| | | | unity-oriented projects | | 2 | | | | | |
| | | | | onnect programs provides b | anafits to | | | | | |
| | | | • | | | | | | | |
| | | | | so that they may feel percepti | • | | | | | |
| | | off p | ost the interaction and | involvement of the Sharda | academic | | | | | |
| | | comm | unity | | | | | | | |
| | | 5. Provi | de ample opportunity | for Sharda University | academic | | | | | |
| | | community to contribute effectively to society and nation building | | | | | | | | |
| 7 | Course | After complet | ion of this course, students | will be able to: | | | | | | |
| | Outcomes | CO1: Students learn to be sensitive to the living challenges of disadvantaged | | | | | | | | |
| | | communities. | | | | | | | | |
| | | CO2: Studen | ts learn to appreciate so | ocietal realities beyond textb | ooks and | | | | | |
| | | classrooms | | | | | | | | |
| | | | | | | | | | | |
| | | CO3: Students learn to apply their knowledge via research, and training for community benefit | | | | | | | | |
| | | CO4: Students learn to work on socio-economic projects with teamwork and timely delivery | | | | | | | | |
| | | CO5: Students society. | s learn to engage with com | munities for meaningful contri | butions to | | | | | |
| | | improve the s | | the gaps and create a plan problems prevailing in differer sinable manner. | | | | | | |



| 8 | Theme | Major research themes: |
|-----|---|---|
| | | Survey and self-learning: In this mode, students will make the survey, analyze data, and will extract results to correlate with their theoretical knowledge. E.g. Crops and animals, land holding, labor problems, medical problems of animals and humans, savage and sanitation situations, waste management, etc. Survey and solution providing: In this mode, students will identify the common problems and will provide solutions/ educate the rural population. E.g. air and water pollution, the need for treatment, use of renewable (mainly solar) energy, electricity-saving devices, inefficiencies in the cropping system, animal husbandry, poultry, pest control, irrigation, machining in agriculture, etc. Survey and reporting: In this mode, students will educate villagers and survey the ground-level status of various government schemes meant for rural development. The analyzed results will be reported to concerned agencies which will help them for taking necessary/corrective measures. E.g. Pradhan Mantri Jan Dhan Yojana, Pradhan Mantri MUDRA Yojana, Pradhan Mantri Jan Dhan Yojana, Pradhan Mantri FasalBima Yojana, Swachh Bharat Abhiyan, Soil Health Card Scheme, Digital India, Skill India Program, BetiBachao, BetiPadhao Yojana, DeenDayal Upadhyaya Gram Jyoti Yojana, Shyama Prasad Mukherjee Rurban Mission, UJWAL Discom Assurance Yojana, Pradhan Mantri Jan Aushadhi Yojana, Pradhan Mantri KhanijKshetra Kalyan Yojana, Pradhan Mantri Suraksha Bima Yojana, Pradhan Mantri Sukanya Samriddhi Yojana, Sansad Adarsh Gram Yojana, Pradhan Mantri Suraksha Bima Yojana, Pradhan Mantri Suraksha Toi yajana, Pradhan Mantri Suraksha Bima Yojana, Pradhan Mantri Suraksha Mantri KojgarProtsahan Yojana, Midday Meal Scheme, Pradhan Mantri Vaya Vandana Yojana, Pradhan Mantri Suraksha Fira Yojana, Pradhan Mantri Suraksha Yojana, Pradhan Mantri SurakshitMatritva Vandana Yojana, and Ayushman Bharat Yojana. |
| 9.1 | <u>Guidelines for</u> <u>Faculty</u> <u>Members</u> | It will be a group assignment. There should be no more than 10 students in each group. The faculty guide will guide the students and approve the project title and help the student in preparing the questionnaire and final report. The questionnaire should be well-designed and it should carry at least 20 questions (Including demographic questions). The faculty will guide the student to prepare the PPT. The topic of the research should be related to social, economical, or environmental issues concerning the common man. The report should contain 2,500 to 3,000 words and relevant charts, tables, and photographs. A plagiarism check of the report must. ETE will conduct out of 100, divided in three parts (i) 30 Marks for the report (ii) 30 Marks for the presentation (iii) 40 Marks for knowledge. The student should submit the report to CCC-Coordinator signed by the faculty guide by The students have to send the hard copy of the report and PPT , and then only |



| | | they will be allowed for ETE. |
|-----|-----------------------------|---|
| 9.2 | Role of CCC- Coordinator | The CCC Coordinator will supervise the whole process and assign students to faculty members. |
| | | 1. UG- B.ScSemester VI - the students will be allocated to faculty member (mentors/faculty member) in odd term. |
| 9.3 | Layout of the Report | Abstract (250 words) a. Introduction b. Literature review(optional) c. Objective of the research d. Research Methodology e. Finding and discussion f. Conclusion and recommendation g. References Note: Research report should base on primary data. |
| 9.4 | Guideline for | Title Page: The following elements must be included: |
| | Report Writing | Title of the article; Name(s) and initial(s) of author(s), preferably with first names spelled out; Affiliation(s) of author(s); Name of the faculty guide and Co-guide Abstract: Each article is to be preceded by a succinct abstract, of up to 250 words, that highlights the objectives, methods, results, and conclusions of the paper. Text: Manuscripts should be submitted in Word. |
| | | Use a normal, plain font (e.g., 12-point Times Roman) for text. Use italics for emphasis. Use the automatic page numbering function to number the pages. Save your file in docx format (Word 2007 or higher) or doc format (older |
| | | Word versions) Reference list: The list of references should only include works that are cited in the text and that have been published or accepted for publication. |
| | | The entries in the list should be in alphabetical order. Journal article Hamburger, C.: Quasimonotonicity, regularity and duality for nonlinear systems of partial differential equations. Ann. Mat. Pura Appl. 169, 321–354 (1995) Article by DOI Sajti, C.L., Georgio, S., Khodorkovsky, V., Marine, W.: New nanohybrid materials for biophotonics. Appl. Phys. A (2007). doi:10.1007/s00339-007- 4137-z Book Geddes, K.O., Czapor, S.R., Labahn, G.: Algorithms for Computer Algebra. |
| | | Kluwer, Boston (1992) Book chapter Broy, M.: Software engineering — from auxiliary to key technologies. In: Broy, M., Denert, E. (eds.) Software Pioneers, pp. 10–13. Springer, Heidelberg (2002) Online document Cartwright, J.: Big stars have weather too. IOP Publishing PhysicsWeb. http://physicsweb.org/articles/news/11/6/16/1 (2007). Accessed 26 June 2007 |



| | | Always use the stand | lard abbreviation of a journal's name according to the ISSN | | | | | | | | |
|-------|---------------|-----------------------|--|--|--|--|--|--|--|--|--|
| | | List of Title Word A | | | | | | | | | |
| | | | 61-LTWA-online.php | | | | | | | | |
| | | | dNote, Springer provides an output style that supports the | | | | | | | | |
| | | | citations and reference list. | | | | | | | | |
| | | EndNote style (zip, 2 | | | | | | | | | |
| | | | re to be numbered using Arabic numerals. | | | | | | | | |
| | | | All figures are to be numbered using Arabic numerals. | | | | | | | | |
| 9.5 | Format: | | oe Spiral/ hardbound | | | | | | | | |
| | | | over page to report will be given by the Coordinator- CCC | | | | | | | | |
| | | Cover page | | | | | | | | | |
| | | Acknowledgement | | | | | | | | | |
| | | Content | | | | | | | | | |
| | | Project report | | | | | | | | | |
| | | Appendices | • | | | | | | | | |
| 9.6 | Important | | are questionnaire and get it approved by concern faculty | | | | | | | | |
| | Dates: | | the final questionnaire withinto CCC- | | | | | | | | |
| | | Coordinator. | 1 | | | | | | | | |
| | | Students will comple | ete their survey work within and submit the same | | | | | | | | |
| | | | ember. (Each group should complete 50 questionnaires) | | | | | | | | |
| | | • | how the 1st draft of the report to concern faculty member | | | | | | | | |
| | | | and submit the same to concern faculty member. | | | | | | | | |
| | | | build give required inputs, so that students can improve their | | | | | | | | |
| | | - | ke the final report submission on | | | | | | | | |
| | | | submit the hard copy and soft copy of the report to CCC- | | | | | | | | |
| | | | by the faculty guide within | | | | | | | | |
| | | • | I submit the soft copy of the PPT to CCC-Coordinator | | | | | | | | |
| | | signed by the facult | | | | | | | | | |
| | | | on will be organized on | | | | | | | | |
| 9.7 | ETE | | e evaluated by panel of faculty members on the basis of | | | | | | | | |
| | | their presentation o | | | | | | | | | |
| | | then presentation o | | | | | | | | | |
| 10 | Course Evalu | ation | | | | | | | | | |
| 10.01 | Continuous A | ssessment | 25% | | | | | | | | |
| | Questionnaire | | | | | | | | | | |
| | Report Writin | | | | | | | | | | |
| 10.02 | ETE (PPT pr | | 75% | | | | | | | | |

| PO | РО | PO | РО | РО | РО | РО | РО | РО | PO | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| CCU108.1 | 1 | 2 | 2 | 3 | 3 | 2 | 3 | 3 | 3 | 2 | 1 | 2 | 2 | 3 |
| CCU108.2 | 1 | 2 | 2 | 3 | 3 | 2 | 3 | 3 | 3 | 2 | 1 | 2 | 2 | 3 |
| CCU108.3 | 1 | 2 | 2 | 3 | 3 | 2 | 3 | 3 | 3 | 2 | 1 | 2 | 2 | 3 |
| CCU108.4 | 1 | 2 | 2 | 3 | 3 | 2 | 3 | 3 | 3 | 2 | 1 | 2 | 2 | 3 |
| CCU108.5 | 1 | 2 | 2 | 3 | 3 | 2 | 3 | 3 | 3 | 2 | 1 | 2 | 2 | 3 |
| CCU108.6 | 1 | 2 | 2 | 3 | 3 | 2 | 3 | 3 | 3 | 2 | 1 | 2 | 2 | 3 |
| Average | 1.0 | 2.0 | 2.0 | 3.0 | 3.0 | 2.0 | 3.0 | 3.0 | 3.0 | 2.0 | 1.0 | 2.0 | 2.0 | 3.0 |



| Sch | ool: SSBSR | Batch: 2024-28 | | | | | | | | | | | | |
|-----------------|-------------------------------|---|----------------|--|--|--|--|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2026-27 | | | | | | | | | | | | |
| (Ho | | | | | | | | | | | | | | |
| | nch: Data Science & lytics | semester: VI | | | | | | | | | | | | |
| <u>Апа</u> 1 | Course Code | CMS371 | | | | | | | | | | | | |
| 2 | Course Title | Numerical Methods Lab | | | | | | | | | | | | |
| 2 | Credits | | | | | | | | | | | | | |
| <u> </u> | | 1 | | | | | | | | | | | | |
| 4 | Contact Hours (L-T-P) | 0-0-2 | | | | | | | | | | | | |
| | Course Status | CC | | | | | | | | | | | | |
| 5 | Course | 1. To provide the student with numerical methods of solving the | a non linear | | | | | | | | | | | |
| 5 | Objective | equations, interpolation, differentiation, and integration. | | | | | | | | | | | | |
| | Objective | 2.To improve the student's skills in numerical methods by | v using the | | | | | | | | | | | |
| | | MATLAB. | 5 | | | | | | | | | | | |
| | | 3. To provide the students are able to formulate a real-world p | problem as a | | | | | | | | | | | |
| | | mathematical programming model, understand the theoretical wo | | | | | | | | | | | | |
| | | simplex method for linear programming and perform iterations o | | | | | | | | | | | | |
| | | relationship between a linear program and its dual, including stron | | | | | | | | | | | | |
| | | complementary slackness and solve specialized linear programming like the transportation and assignment problems. | ing problems | | | | | | | | | | | |
| 6 | Course | CO1: Understand the procedures, algorithms, and concepts require | re tosolve | | | | | | | | | | | |
| 0 | Outcomes | specific problems. | | | | | | | | | | | | |
| | | CO2: Discuss and develop the algorithms to solve system of tr | anscendental | | | | | | | | | | | |
| | | equations and measure the accuracy. | | | | | | | | | | | | |
| | | CO3: Discuss and develop the algorithms to solve finite difference | s and | | | | | | | | | | | |
| | | interpolation and measure the accuracy. | _ | | | | | | | | | | | |
| | | CO4: Discuss and develop the algorithms to solve divided differen | nces and | | | | | | | | | | | |
| | | measure the accuracy. | | | | | | | | | | | | |
| | | CO5: Discuss and develop the algorithms to solve numerical differences measure the accuracy. | rentiation and | | | | | | | | | | | |
| | | CO6: Discuss and develop the algorithms to solve numerical in | tegration and | | | | | | | | | | | |
| | | measure the accuracy. | - | | | | | | | | | | | |
| 7 | Course | This course is an introduction to the numerical analysis. Thepri | mary | | | | | | | | | | | |
| | Description | objective of the course is to develop the basic | loomithmag to | | | | | | | | | | | |
| | | understanding of numerical algorithms and skills to implemental solve mathematical problems in MATLAB. | ligorithins to | | | | | | | | | | | |
| 8 | Outline syllabus | solve mathematical problems in WATLAD. | СО | | | | | | | | | | | |
| Ũ | • | | Mapping | | | | | | | | | | | |
| | Unit 1 | | | | | | | | | | | | | |
| | A, B, C | Solution of system of linear equations Cramer's rule | CO1 | | | | | | | | | | | |
| | | i) Cramer's ruleii) Gauss elimination and Gauss-Jordan method | | | | | | | | | | | | |
| | | iii) Jacobi's method, Gauss-Seidal method. | | | | | | | | | | | | |
| | Unit 2 | | | | | | | | | | | | | |
| | A, B, C | 2. System of Transcendental equations | CO2 | | | | | | | | | | | |
| | | i) Bisection method and Method of false position | | | | | | | | | | | | |
| | | ii)Secant method, iteration method | | | | | | | | | | | | |
| | Un:4 2 | iii) Newton-Raphson method | | | | | | | | | | | | |
| | Unit 3 | 2 Einite differences and Internal-tion | CO3 | | | | | | | | | | | |
| | A, B, C | 3. Finite differences and Interpolation i) Newton's forward, backward and divided difference | CO3 | | | | | | | | | | | |
| | | interpolations | | | | | | | | | | | | |
| | Unit 4 | | | | | | | | | | | | | |
| | A, B, C | 4. Divided differences | CO4 | | | | | | | | | | | |



| | i) Newton's divided difference formulaii) Lagrange's interpolation formula. | |
|---------------------------|---|---------|
| Unit 5 | n) Lagrange's interpolation formula. | |
| A, B, C | 5.Numerical differentiation and integration i) Newton's forward and backward formula ii) Trapezoidal rule and Simpson's 1/3 and 3/8 rules. | CO5,CO6 |
| Mode of examination | Practical | |
| Weightage Distribution | CA:25%; CE:25%; ESE:50% | |
| Text book/s* | Gerald Recktenwald: Introduction to Numerical Methods, CRC Press. John H. Mathews, Pearson: Numerical Methods Using MATLAB, | |
| Other References | César Pérez López: MATLAB Programming for Numerical Analysis, Apress. Steven T., Karris, Numerical Analysis: Using Matlab And Excel, Orchard Publications. | |

| РО | РО | PO | РО | РО | PO | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| CMS371.1 | 3 | 3 | 3 | 2 | 2 | 1 | 2 | 3 | 1 | 1 | 3 | | 1 | |
| CMS371.2 | 3 | 3 | 3 | 2 | 2 | 1 | 2 | 3 | 1 | 1 | 3 | | 1 | |
| CMS371.3 | 3 | 3 | 3 | 2 | 2 | 1 | 2 | 3 | 1 | 1 | 3 | | 1 | |
| CMS371.4 | 3 | 3 | 3 | 2 | 2 | 1 | 2 | 3 | 1 | 1 | 3 | | 1 | |
| CMS371.5 | 3 | 3 | 3 | 2 | 2 | 1 | 2 | 3 | 1 | 1 | 3 | | 1 | |
| CMS371.6 | 3 | 3 | 3 | 2 | 2 | 1 | 2 | 3 | 1 | 1 | 3 | | 1 | |
| Average | 3.0 | 3.0 | 3.0 | 2.0 | 2.0 | 1.0 | 2.0 | 3.0 | 1.0 | 1.0 | 3.0 | | 1.0 | |



| Scho | ol: SSBSR | Batch: 2024-28 | | | | | | | | | | | | |
|------|--------------------------|--|---------------|--|--|--|--|--|--|--|--|--|--|--|
| Prog | ramme: B.Sc. | Academic Year: 2026-27 | | | | | | | | | | | | |
| (Hor | | | | | | | | | | | | | | |
| | ich: Data Science | Semester: VI | | | | | | | | | | | | |
| | nalytics | | | | | | | | | | | | | |
| 1 | Course Code | BDA360 | | | | | | | | | | | | |
| 2 | Course Title | Statistical Simulation Lab | | | | | | | | | | | | |
| 3 | Credits | 1 | | | | | | | | | | | | |
| 4 | Contact Hours (L-T-P) | 0-0-2 | | | | | | | | | | | | |
| | Course Status | СС | | | | | | | | | | | | |
| 5 | Course | The learning objectives include: Concept of simulation and simulation | on modeling | | | | | | | | | | | |
| 5 | Objective | Generation of Pseudo random number generators as well as from | | | | | | | | | | | | |
| | Objective | statistical distributions, Monte-Carlo simulation technique and ar | | | | | | | | | | | | |
| | | simulation techniques. | pheation of | | | | | | | | | | | |
| 6 | Course | CO1: Recognize the concepts of probability and statistics that are relev | ant to | | | | | | | | | | | |
| ÷ | Outcomes | modeling and simulation (K2, K3). | | | | | | | | | | | | |
| | | CO2: How to generate random numbers by the different methods (K2, | K3). | | | | | | | | | | | |
| | | CO3: Design and implement Bootstrapping; jackknife resampling (K3, | | | | | | | | | | | | |
| | | CO4: Be able to evaluate and interpret the Markov-Chain Monte Ca | rlo (MCMC) | | | | | | | | | | | |
| | | simulations (K3, K4). | | | | | | | | | | | | |
| | | CO5: Hands-on experience in using simulation software packag | es/structured | | | | | | | | | | | |
| | | programming languages (K3, K4, K5) | | | | | | | | | | | | |
| | | CO6: How simulation may be used to understand the behavior of | | | | | | | | | | | | |
| | | systems by utilizing mathematical models with an emphasis on sim | ulation (K4, | | | | | | | | | | | |
| | | Кб). | | | | | | | | | | | | |
| 7 | Course | The course topics will include a review of concepts from probability and st | | | | | | | | | | | | |
| | Description | that are relevant to modeling and simulation, algorithms for rand | | | | | | | | | | | | |
| | | sampling, modeling and analysis of basic queueing systems, variar | | | | | | | | | | | | |
| | | techniques, statistical-validation techniques, Independent Monte Carl | | | | | | | | | | | | |
| | | Markov-Chain Monte Carlo (MCMC) simulations, and discrete-eve and simulation. | ent modeling | | | | | | | | | | | |
| 0 | Outline avillation | | СО | | | | | | | | | | | |
| 8 | Outline syllabus | | Mapping | | | | | | | | | | | |
| | Unit 1 | Lab. Experiment 1 | / | | | | | | | | | | | |
| | A, B, C | Review of R/Python. | CO1 | | | | | | | | | | | |
| | | Problem Based on Random number generation: Inverse-transform; | CO1 | | | | | | | | | | | |
| | | acceptance-rejection; transformations. | | | | | | | | | | | | |
| | | Problem Based on Statistic simulations: generating random variables, | CO1 | | | | | | | | | | | |
| | | simulating normal, gamma and beta random variables. | | | | | | | | | | | | |
| | Unit 2 | Lab. Experiment 2 | | | | | | | | | | | | |
| | A, B, C | Problem Based on Simulating multivariate distributions, MCMC | CO2 | | | | | | | | | | | |
| | , , | methods. | | | | | | | | | | | | |
| | | Problem Based on Gibbs sampler, simulating random fields, | CO2 | | | | | | | | | | | |
| | | simulating stochastic process. | | | | | | | | | | | | |
| | | Problem Based on Variance reduction technique: importance | CO2 | | | | | | | | | | | |
| | | sampling for integration, control variates and antithetic variables. | | | | | | | | | | | | |
| | Unit 3 | Lab. Experiment 2 | | | | | | | | | | | | |
| | A, B, C | Problem Based on Bootstrapping; jackknife resampling. | CO3 | | | | | | | | | | | |
| | , , | Problem Based on Bootstrapping for estimation of sampling | CO3 | | | | | | | | | | | |
| | | distribution. Confidence intervals, variance stabilizing transformation. | | | | | | | | | | | | |
| | | Problem Based on Bootstrapping in regression and sampling from | CO3 | | | | | | | | | | | |
| | | FIGURE DASCU ON DOUGUADDING IN TEGRESSION and Samoning norm | | | | | | | | | | | | |



| Unit 4 | Lab. Experiment 2 | |
|---------------------------|--|----------|
| A, B, C | Problem Based on Simulating a non-homogeneous Poisson process. | CO4 |
| | Problem Based on Optimization using Monte Carlo methods simulated annealing for optimization. | CO4 |
| | Problem Based on Solving differential equations by Monte Carlo methods. | CO4 |
| Unit 5 | Lab. Experiment 2 | |
| A, B, C | Problem Based on Univariate density estimation; kernel smoothing multivariate density estimation | CO5, CO6 |
| | Problem Based on Root finding; more on numerical integration; numerical maximization/minimization; constrained and unconstrained optimization. | |
| | Problem Based on EM (Expectation-Maximization) algorithm; simplex algorithm. | CO5, CO6 |
| Mode of examination | Practical+Viva | |
| Weightage Distribution | CA:25%; CE:25%; ESE:50% | |
| Text book/s* | ¹ I.Fishman, G.S. Monte Carlo: Concepts, Algorithms and Applications. 2.Rubinstein, R.Y.: Simulation and the Monte Carlo Method. | |
| Other | 1. Ross, S. M.: Simulation, Third Edition, Academic Press. | |
| References | 2. Efron, B. and Tibshirani. R.J.: An introduction to the Bootstrap. | |

| РО | РО | PO | РО | PO | PO | PO | PO | PO | PO | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA360.1 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| BDA360.2 | 1 | 2 | 3 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| BDA360.3 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| BDA360.4 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| BDA360.5 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| BDA360.6 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 3.0 | 1.0 | 2.0 | 3.0 |



| | ool: SSBSR | Batch: 2024-28 | | | | | | | | | | |
|-----|--|--|---|--|--|--|--|--|--|--|--|--|
| (Ho | | Academic Year: 2026-27 | | | | | | | | | | |
| Ana | nch: Data Science & lytics | | | | | | | | | | | |
| 1 | Course Code | BDA361 | | | | | | | | | | |
| 2 | Course Title | Multivariate Data Analysis Lab | | | | | | | | | | |
| 3 | Credits | 1 | | | | | | | | | | |
| 4 | Contact Hours(L- T-P) | 0-0-2 | | | | | | | | | | |
| | Course Status | CC | | | | | | | | | | |
| 5 | Course Familiarise students with the multivariate normal distribution, est Objective the mean vector and the covariance matrix, the distributions and the sample correlation coefficients, classification of observations, the distribution of the sample covariance matrix, and the sample gene variance. Course Course | | | | | | | | | | | |
| 6 | Course Outcomes | CO1: Demonstrate knowledge and understanding of the multivariate normal distribution. (K2, K3) CO2: Demonstrate knowledge and understanding of the concept of estimation of the mean vector and the covariance matrix. (K2, K3) CO3: Demonstrate advanced understanding of the concepts of dimension reduction technique. (K2, K3) CO4: Describe the concepts of how to use and apply dependence techniques in multivariate data analysis. (K2, K3) CO5: Describe the concepts of analysis of variance and covariance in multivariate data analysis. (K3, K4, K5) | | | | | | | | | | |
| | | CO6: Apply the statistical tool and software in multivariate data a | nalysis. (K2, | | | | | | | | | |
| 7 | Course Description | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, and | riate normal matrix, the sification of | | | | | | | | | |
| 7 8 | Description Outline syllabus | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, class | riate normal matrix, the sification of | | | | | | | | | |
| | Description Outline syllabus Unit 1 | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, an generalized variance. | riate normal matrix, the sification of d the sample CO Mapping | | | | | | | | | |
| | Description Outline syllabus | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, an generalized variance. | riate normal matrix, the sification of d the sample CO Mapping CO1 | | | | | | | | | |
| | Description Outline syllabus Unit 1 | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, an generalized variance. | riate normal matrix, the sification of d the sample CO Mapping CO1 CO1 | | | | | | | | | |
| | Description Outline syllabus Unit 1 A, B, C | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, an generalized variance. | riate normal matrix, the sification of d the sample CO Mapping CO1 | | | | | | | | | |
| | Description Description Unit 1 A, B, C Unit 2 Unit 2 | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, an generalized variance. Problem based on Data Cleaning and Data Screening Problem based on to check Data Normality Problem based on to check Reliability Testing | riate normal matrix, the sification of d the sample CO Mapping CO1 CO1 CO1 | | | | | | | | | |
| | Description Outline syllabus Unit 1 A, B, C | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, and generalized variance. Problem based on Data Cleaning and Data Screening Problem based on to check Data Normality Problem based on to check Reliability Testing Problem based on Multiple and Partial correlation | riate normal matrix, the sification of d the sample CO Mapping CO1 CO1 CO1 CO1 | | | | | | | | | |
| | Description Outline syllabus Unit 1 A, B, C Unit 2 | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, an generalized variance. Problem based on Data Cleaning and Data Screening Problem based on to check Data Normality Problem based on to check Reliability Testing | riate normal matrix, the sification of d the sample CO1 CO1 CO1 CO1 CO1 CO2 CO2 | | | | | | | | | |
| | Description Description Unit 1 A, B, C Unit 2 A, B, C | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, and generalized variance. Problem based on Data Cleaning and Data Screening Problem based on to check Data Normality Problem based on to check Reliability Testing Problem based on Multiple and Partial correlation | riate normal matrix, the sification of d the sample CO Mapping CO1 CO1 CO1 CO1 | | | | | | | | | |
| | Description Description Unit 1 A, B, C Unit 2 A, B, C Unit 3 Unit 3 | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, an generalized variance. Problem based on Data Cleaning and Data Screening Problem based on to check Data Normality Problem based on to check Reliability Testing Problem based on Multiple and Partial correlation Problem based on Canonical correlation | riate normal matrix, the sification of d the sample CO Mapping CO1 CO1 CO1 CO1 CO2 CO2 CO2 CO2 | | | | | | | | | |
| | Description Description Unit 1 A, B, C Unit 2 A, B, C | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, and generalized variance. Problem based on Data Cleaning and Data Screening Problem based on to check Data Normality Problem based on to check Reliability Testing Problem based on Multiple and Partial correlation Problem based on Canonical correlation Problem based on Principal Component Analysis | riate normal matrix, the sification of d the sample CO1 CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO2 | | | | | | | | | |
| | Description Description Unit 1 A, B, C Unit 2 A, B, C Unit 3 Unit 3 | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, an generalized variance. Problem based on Data Cleaning and Data Screening Problem based on to check Data Normality Problem based on to check Reliability Testing Problem based on Multiple and Partial correlation Problem based on Canonical correlation | riate normal matrix, the sification of d the sample CO Mapping CO1 CO1 CO1 CO1 CO2 CO2 CO2 CO2 | | | | | | | | | |
| | Description Description Unit 1 A, B, C Unit 2 A, B, C Unit 3 Unit 3 | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, an generalized variance. Problem based on Data Cleaning and Data Screening Problem based on to check Data Normality Problem based on to check Reliability Testing Problem based on Multiple and Partial correlation Problem based on Canonical correlation Problem based on Principal Component Analysis Problem based on Cluster Analysis: Exploratory factor analysis Problem based on Cluster Analysis: Hierarchal Cluster and Non- | riate normal matrix, the sification of d the sample CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO2 CO2 | | | | | | | | | |
| | Description Description Outline syllabus Unit 1 A, B, C Unit 2 A, B, C Unit 3 A, B, C | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, an generalized variance. Problem based on Data Cleaning and Data Screening Problem based on to check Data Normality Problem based on to check Reliability Testing Problem based on Multiple and Partial correlation Problem based on Canonical correlation Problem based on Principal Component Analysis Problem based on Cluster Analysis: Exploratory factor analysis Problem based on Cluster Analysis: Hierarchal Cluster and Non- | riate normal matrix, the sification of d the sample CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO2 CO2 | | | | | | | | | |
| | Description Description Outline syllabus Unit 1 A, B, C Unit 2 A, B, C Unit 3 A, B, C Unit 3 A, B, C Unit 4 | CO6: Apply the statistical tool and software in multivariate data a K6) This module aims to provide an understanding of the multiva distribution, estimation of the mean vector and the covariance distributions and uses of sample correlation coefficients, clas observations, the distribution of the sample covariance matrix, an generalized variance. Problem based on Data Cleaning and Data Screening Problem based on to check Data Normality Problem based on to check Reliability Testing Problem based on Multiple and Partial correlation Problem based on Canonical correlation Problem based on Principal Component Analysis Problem based on Cluster Analysis: Hierarchal Cluster and Nonhierarchal Cluster | riate normal matrix, the sification of d the sample CO Mapping CO1 CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO2 CO3 CO3 CO3 | | | | | | | | | |



| Unit 5 | | |
|--------------|---|----------|
| A, B, C | Problem based on Analysis of Variance | CO5, CO6 |
| | Problem based on Analysis of and Covariance | CO5, CO6 |
| | Problem based on Multivariate Analysis of Variance and Covariance | CO5, CO6 |
| Mode of | Practical+Viva | |
| examination | | |
| Weightage | CA:25%; CE:25%; ESE:50% | |
| Distribution | CA.2576, CE.2576, ESE.5676 | |
| Text book/s* | 1. Johnson, R.A. and Wichern, D.W.: Applied Multivariate | |
| | Statistical Analysis, Sixth Edition, Pearson Education India. | |
| | 2.Hardle, W.K. and Hlavka, Z. Multivariate Statistics, Springer. | |
| Other | 1. Anderson, T.W.: An Introduction to Multivariate Statistical | |
| References | Analysis, Third Edition, Wiley. | |
| | 2.Härdle, W.K. and Simar, L. : Applied Multivariate Statistical | |
| | Analysis, Springer. | |

| РО | PO | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| BDA361.1 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| BDA361.2 | 1 | 2 | 3 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| BDA361.3 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| BDA361.4 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| BDA361.5 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| BDA361.6 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 3.0 | 1.0 | 2.0 | 3.0 |



Detailed Syllabus for

HONOURS

OR

HONOURS WITH RESEARCH

IN

DATA SCIENCE & ANLYTICS



| | ool: SSBSR | Batch: 2024-28 | | | | | | | | | | |
|-----|---|--|--|--|--|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2027-28 | | | | | | | | | | |
| (Ho | ns.) nch: Data Science | | | | | | | | | | | |
| | nch: Data Science | Semester: VII | | | | | | | | | | |
| 1 | Course Code | MDA104 | | | | | | | | | | |
| 2 | Course Title | Next Generation Databases | | | | | | | | | | |
| 3 | Credits | | | | | | | | | | | |
| 4 | Contact Hours | 4 4-0-0 | | | | | | | | | | |
| 4 | (L-T-P) | 4-0-0 | | | | | | | | | | |
| | Course Status | CC | | | | | | | | | | |
| 5 | Course | To explore the concepts of NoSQL Databases. To understand and | d use | | | | | | | | | |
| 5 | Objective | columnar and distributed database patterns. | | | | | | | | | | |
| 6 | Course | After completion of this course, students will be able to | | | | | | | | | | |
| U | Outcomes | CO1: Develop and Explore the relationship between Big-Data and | NoSOL | | | | | | | | | |
| | | databases. (K1, K2, K3) | X | | | | | | | | | |
| | | CO2: Formulate a fundamental relationship between Big-Data and N | NoSQL | | | | | | | | | |
| | | databases. (K2, K3) | | | | | | | | | | |
| | | CO3: Describe various types of NoSQL databases to analyze the big | data for | | | | | | | | | |
| | | useful business applications. (K3, K4) CO4: Derive and Work with NoSQL databases to analyze the big | datafar | | | | | | | | | |
| | | useful business applications. (K4, K5) | , ualator | | | | | | | | | |
| | | CO5: Discuss different data models to suit various data represent | tations and | | | | | | | | | |
| | | - | | | | | | | | | | |
| | | storage needs. (K5, K6) CO6: Explain and correlate with different data models to suit variousdata | | | | | | | | | | |
| | | | riousdata | | | | | | | | | |
| | | CO6: Explain and correlate with different data models to suit va representations and storage needs. (K5, K6) | nriousdata | | | | | | | | | |
| 7 | Course | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) | | | | | | | | | | |
| 7 | Course Description | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6)To integrate the intrinsic ideas for the use of various Data models | | | | | | | | | | |
| 7 | | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) | | | | | | | | | | |
| 7 | | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6)To integrate the intrinsic ideas for the use of various Data models of databases. | for a variety | | | | | | | | | |
| - | Description Outline syllabus | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6)To integrate the intrinsic ideas for the use of various Data models of databases. | for a variety | | | | | | | | | |
| - | Description Outline syllabus Unit 1 | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6)To integrate the intrinsic ideas for the use of various Data models of databases. | for a variety CO Mapping | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. | for a variety CO Mapping CO1 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. | for a variety CO Mapping CO1 CO1 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. | for a variety CO Mapping CO1 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. | for a variety CO Mapping CO1 CO1 CO1 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. | for a variety CO Mapping CO1 CO1 CO1 CO2 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. Birth of NoSQL-Document Database—XML Databases. | for a variety CO Mapping CO1 CO1 CO1 CO2 CO2 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C C | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. | for a variety CO Mapping CO1 CO1 CO1 CO2 | | | | | | | | | |
| | Description Outline syllabus Unit 1 A B C Unit 2 A B C C Unit 2 Unit 3 | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. Birth of NoSQL-Document Database—XML Databases. JSON Document Database-Graph Databases. | for a variety CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C C | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. Birth of NoSQL-Document Database—XML Databases. JSON Document Databases-Graph Databases. | for a variety CO Mapping CO1 CO1 CO1 CO2 CO2 | | | | | | | | | |
| | Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 3 A | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. Birth of NoSQL-Document Database—XML Databases. JSON Document Databases-Graph Databases. ColumnDatabases—Data Warehousing Schemes-Columnar Alternative-Sybase IQ-C-Store. | for a variety CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO3 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C C Unit 2 Unit 3 | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. Big-Data Revolution-CAP Theorem. Birth of NoSQL-Document Database—XML Databases. JSON Document Databases-Graph Databases. ColumnDatabases—Data Warehousing Schemes-Columnar Alternative-Sybase IQ-C-Store. Vertica-Column Database Architectures-SSD and In-Memory | for a variety CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 2 A B C Unit 3 A B B | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. Birth of NoSQL-Document Database—XML Databases. JSON Document Databases-Graph Databases. ColumnDatabases—Data Warehousing Schemes-Columnar Alternative-Sybase IQ-C-Store. Vertica-Column Database Architectures-SSD and In-Memory Databases. | for a variety CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO3 CO3 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 3 A B C C C C | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. Big-Data Revolution-CAP Theorem. Birth of NoSQL-Document Database—XML Databases. JSON Document Databases-Graph Databases. ColumnDatabases—Data Warehousing Schemes-Columnar Alternative-Sybase IQ-C-Store. Vertica-Column Database Architectures-SSD and In-Memory | for a variety CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO3 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 3 A B C C Unit 3 C Unit 4 | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. Birth of NoSQL-Document Database—XML Databases. JSON Document Databases-Graph Databases. JSON Document Databases-Graph Databases. Vertica-Column Database Architectures-SSD and In-Memory Databases. In-Memory Databases-Berkeley Analytics Data Stack andSpark. | for a variety CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO3 CO3 CO3 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 3 A B C C C C | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. Birth of NoSQL-Document Database—XML Databases. JSON Document Database-Storage Schemes-Columnar Alternative-Sybase IQ-C-Store. Vertica-Column Database Architectures-SSD and In-Memory Databases. In-Memory Databases-Berkeley Analytics Data Stack andSpark. Distributed Database Patterns— Distributed Relational Databases- | for a variety CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO3 CO3 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 3 A B C Unit 3 A B C Unit 4 A | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. Birth of NoSQL-Document Database—XML Databases. JSON Document Database—XML Databases. JSON Document Databases-Graph Databases. Vertica-Column Database Architectures-SSD and In-Memory Databases. In-Memory Databases-Berkeley Analytics Data Stack andSpark. Distributed Database Patterns— Distributed Relational Databases-Non- relational Distributed Databases. | for a variety CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO3 CO3 CO3 CO3 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 3 A B C C Unit 3 C Unit 4 | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. Birth of NoSQL-Document Database—XML Databases. JSON Document Database-Tanabase. ColumnDatabases—Data Warehousing Schemes-Columnar Alternative-Sybase IQ-C-Store. Vertica-Column Database Architectures-SSD and In-Memory Databases. In-Memory Databases-Berkeley Analytics Data Stack andSpark. Distributed Database Patterns— Distributed Relational Databases- Non- relational Distributed Databases. | for a variety CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO2 CO3 CO3 CO3 | | | | | | | | | |
| - | Description Outline syllabus Unit 1 A B C Unit 2 A B C Unit 3 A B C Unit 3 A B C Unit 4 A | CO6: Explain and correlate with different data models to suit varepresentations and storage needs. (K5, K6) To integrate the intrinsic ideas for the use of various Data models of databases. Database Revolutions- system Architecture-RelationalDatabase. Database Design-Data Storage-Transaction Management. Data warehouse and Data Mining-Information Retrieval. Big-Data Revolution-CAP Theorem. Birth of NoSQL-Document Database—XML Databases. JSON Document Database—XML Databases. JSON Document Databases-Graph Databases. Vertica-Column Database Architectures-SSD and In-Memory Databases. In-Memory Databases-Berkeley Analytics Data Stack andSpark. Distributed Database Patterns— Distributed Relational Databases-Non- relational Distributed Databases. | for a variety CO Mapping CO1 CO1 CO1 CO2 CO2 CO2 CO3 CO3 CO3 CO3 | | | | | | | | | |



| | Cassandra Consistency. | | | | | | | |
|---------------------------|---|----------|--|--|--|--|--|--|
| Unit 5 | | | | | | | | |
| A | Data Models and Storage-SQL-NoSQLAP Is-Return SQL-Advance Databases—Postgre SQL. | CO5, CO6 | | | | | | |
| В | Riak-CouchDB-NEO4J-Redis-Future, Databases- Revolution Revisited-Counter revolutionaries-Oracle HQ. | | | | | | | |
| С | Other Convergent Databases-Disruptive Database Technologies. | CO5, CO6 | | | | | | |
| Mode of examination | Theory | | | | | | | |
| Weightage Distribution | CA:25%; ESE:75% | | | | | | | |
| Text book/s* | Abraham Silberschatz, Henry F. Korth, S.Sudarshan, "Database System Concepts", Sixth Edition, McGraw Hill. Guy Harrison, "Next Generation Databases", A Press. | | | | | | | |
| Other References | Eric Redmond, Jim R Wilson, "Seven Databasesin Seven Weeks", LLC. Dan Sullivan, "NoSQL for Mere Mortals", Addison-Wesley, 2015. | | | | | | | |

| РО | PO | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA104.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA104.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA104.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 3 | | 3 | |
| MDA104.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA104.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA104.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 3 | | 3 | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 3.0 | | 3.0 | |



| | ool: SSBSR | Batch: 2024-28 | | | | | | | | | | | | | |
|-----|-------------------------------|--|---------------|--|--|--|--|--|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2027-28 | | | | | | | | | | | | | |
| (Ho | ns.) nch: Data Science | | | | | | | | | | | | | | |
| | nch: Data Science nalytics | Semester: VII | | | | | | | | | | | | | |
| 1 | Course Code | MDA109 | | | | | | | | | | | | | |
| 2 | Course Title | Big Data Analytics | | | | | | | | | | | | | |
| 3 | Credits | | | | | | | | | | | | | | |
| 4 | Contact Hours | | | | | | | | | | | | | | |
| 4 | | 4-0-0 | | | | | | | | | | | | | |
| | (L-T-P) Course Status | CC | | | | | | | | | | | | | |
| 5 | Course | This course is aimed to provide an advanced understanding of big da | data | | | | | | | | | | | | |
| 5 | Objective | | lla | | | | | | | | | | | | |
| (| 5 | overview, model building, clustering, and advanced analytics. CO1: Discuss the concept of big data analysis and data preparation (K3) | | | | | | | | | | | | | |
| 6 | Course Outcomes | | | | | | | | | | | | | | |
| | Outcomes | CO2: Describe the concept model building, communicate results, and chec | | | | | | | | | | | | | |
| | | pasic data analysis. (K1, K2, K3). CO 3: Explain the concept how using R to look at data introduction to 1 | | | | | | | | | | | | | |
| | | Analyzing and Exploring the Data, Statistics for Model Building and | | | | | | | | | | | | | |
| | | Advanced Analytics. (K3, K4) | | | | | | | | | | | | | |
| | | CO 4: Illustrate the concept of K Means Clustering, association | rules, linear | | | | | | | | | | | | |
| | | regression, logistic regression, and Naïve Bayesian Classifier a | | | | | | | | | | | | | |
| | | decision trees, time series analysis, and text analysis. (K2, K3, K4). | | | | | | | | | | | | | |
| | | CO 5: Discuss the concept of unstructured data – Map Reduce and I | | | | | | | | | | | | | |
| | | Hadoop Ecosystem In-database Analytics and illustrate SQL | | | | | | | | | | | | | |
| | | Advanced SQL, and MADlib for In-database Analytics (K3, K4, K5). | | | | | | | | | | | | | |
| | | CO6: Demonstrate the understanding of the Endgame, or putting it | all together: | | | | | | | | | | | | |
| | | operationalizing an analytics project, creating the final deliver visualization techniques, and final lab exercise on big data analytic | | | | | | | | | | | | | |
| | | K6). | | | | | | | | | | | | | |
| 7 | Course | This course is given the deep knowledge of big data, model buildin | g, clustering | | | | | | | | | | | | |
| | Description | and advance analytics. | | | | | | | | | | | | | |
| 8 | Outline syllabus | | CO | | | | | | | | | | | | |
| | Unit 1 | | Mapping | | | | | | | | | | | | |
| | | State of the Practice in Analytics, the Data Scientist, | CO1 | | | | | | | | | | | | |
| | A B | Big Data Analytics in Industry Verticals | CO1 | | | | | | | | | | | | |
| | C | Data Analytics Life cycle: Discovery, Data Preparation, Model | C01 | | | | | | | | | | | | |
| | C | Planning. | 001 | | | | | | | | | | | | |
| | | | | | | | | | | | | | | | |
| | Unit 2 | Model Building, Communicating Results, Operationalizing Review | CO2 | | | | | | | | | | | | |
| | Δ | of Basic Data Analytic Methods Using R: Using R to Look at Data Introduction to R, | CO2 | | | | | | | | | | | | |
| | AB | Analyzing and Exploring the Data, Statistics for Model Building, | CO2 | | | | | | | | | | | | |
| | В | and Evaluation Advanced Analytics. | 02 | | | | | | | | | | | | |
| | С | | | | | | | | | | | | | | |
| | Unit 3 | | | | | | | | | | | | | | |
| | А | K Means Clustering, Association Rules, Linear Regression, | CO3 | | | | | | | | | | | | |
| | В | Logistic Regression, Naïve Bayesian Classifier, | CO3 | | | | | | | | | | | | |
| | С | Decision Trees Time Series Analysis, Text Analysis. | CO3 | | | | | | | | | | | | |
| | Unit 4 | | | | | | | | | | | | | | |
| | A | Technologies and Tools: Analytics for Unstructured Data – Map | CO4 | | | | | | | | | | | | |
| | | Reduce and Hadoop, | | | | | | | | | | | | | |
| | D | The Hadoop Ecosystem In-database Analytics – SQL Essentials | CO4 | | | | | | | | | | | | |
| | В | The Hudoop Deosystem in duduouse Thiarytes SQL Essentials | 004 | | | | | | | | | | | | |



| Unit 5 | | | | | | | | |
|---------|--------------|---|----------|--|--|--|--|--|
| А | A | The Endgame, or Putting it All Together: Operationalizing an Analytics Project, | CO5, CO6 | | | | | |
| В | C | Creating the Final Deliverables, Data Visualization Techniques, | CO5, CO6 | | | | | |
| С | F | Final Lab Exercise on Big Data Analytics. | CO5, CO6 | | | | | |
| Mode | of 7 | Гheory | | | | | | |
| examir | ation | | | | | | | |
| Weight | tage | CA:25%; ESE:75% | | | | | | |
| Distrib | ution | CA:2570; ESE:7570 | | | | | | |
| Text be | ook/s^* 1. | Tom White, "Hadoop: The Definitive Guide", 3rd edition, | | | | | | |
| | 0' | 'Reilly Media. | | | | | | |
| | 2. | Big Data Black Book, Wiley Publications. | | | | | | |
| Other | | V. Prajapati, "Big Data Analytics with R and Hadoop", Packt Pub. | | | | | | |
| Refere | nces 2.1 | N. Dasgupta, Practical Big Data Analytics, Packt Publication | | | | | | |
| | Lt | td. | | | | | | |

| РО | PO | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA109.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA109.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA109.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 3 | | 3 | |
| MDA109.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA109.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA109.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 3 | | 3 | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 3.0 | | 3.0 | |



| Scho | ol: SSBSR | Batch: 2024-28 | | | | | | | | | | | | |
|------|--------------------------|--|--|--|--|--|--|--|--|--|--|--|--|--|
| Prog | ramme: B.Sc. | Academic Year: 2027-28 | | | | | | | | | | | | |
| (Hon | | | | | | | | | | | | | | |
| | ch: Data Science | Semester: VII | | | | | | | | | | | | |
| - | nalytics | | | | | | | | | | | | | |
| 1 | Course Code | MDA110 | | | | | | | | | | | | |
| 2 | Course Title | Time Series, Forecasting and Index Number | | | | | | | | | | | | |
| 3 | Credits | 3 | | | | | | | | | | | | |
| 4 | Contact Hours (L-T-P) | 3-0-0 | | | | | | | | | | | | |
| | Course Status | DSE | | | | | | | | | | | | |
| 5 | Course | The objective of the course is to explain basic concepts of regression, | time | | | | | | | | | | | |
| | Objective | series, forecasting, and index numbers. | | | | | | | | | | | | |
| 6 | Course Outcomes | CO1: Explain and illustrate the nature and uses of forecasts, some time series, the forecasting process, resources for forecastin background for forecasting: graphical displays, numerical descrip series data (K2, K3) CO2: Describe how to evaluate least squares estimation in linea models, statistical inference in linear regression, prediction of new of model adequacy checking, model adequacy checking, generalized a least squares, and regression models for general time series data. (K6) CO3: Explain and illustrate first-order exponential smoothing, mo series data, second-order exponential smoothing, and higher-order smoothing. (K3, K6) CO4: Use forecasting: constant process, linear trend process, and estimation of σe^2 , adaptive updating of the discount factor, assessment. (K3, K6) CO5: Describe autoregressive integrated moving average (ARIMA) m CO6: Explain and illustrate index numbers with the application. (K6) | exponential evaluate the and model | | | | | | | | | | | |
| 7 | Course | This course will cover the fundamental concepts of Regression, | time series, | | | | | | | | | | | |
| 0 | Description | forecasting, and Index numbers. | СО | | | | | | | | | | | |
| 8 | Outline syllabus | | Mapping | | | | | | | | | | | |
| | Unit 1 | | | | | | | | | | | | | |
| | A | Introduction to Forecasting: The Nature and Uses of Forecasts, Some Examples of Time Series, The Forecasting Process, Resources for Forecasting, | CO1 | | | | | | | | | | | |
| | В | Statistics Background for Forecasting: Graphical Displays, Numerical Description of Time Series Data, Use of Data Transformations and Adjustments, | CO1 | | | | | | | | | | | |
| | С | General Approach to Time Series Modeling and Forecasting, Evaluating and Monitoring Forecasting Model Performance | CO1 | | | | | | | | | | | |
| | Unit 2 | | | | | | | | | | | | | |
| | А | Regression Analysis and Forecasting: Least Squares Estimation in Linear Regression Models | CO2 | | | | | | | | | | | |
| | В | Model Adequacy Checking, Generalized and Weighted Least Squares, Regression Models for General Time Series Data. | CO2 | | | | | | | | | | | |
| | С | Statistical Inference in Linear Regression, Prediction of New Observations | CO2 | | | | | | | | | | | |
| | Unit 3 | | | | | | | | | | | | | |
| | А | Introduction of Time series, Utility of Time series, Components of time series, Models of time series, | CO3 | | | | | | | | | | | |
| | В | Methods of measuring linear trends, | CO4 | | | | | | | | | | | |
| | С | Methods of measuring seasonal variation, Method of measuring cyclic variation | CO4 | | | | | | | | | | | |
| | Unit 4 | | | | | | | | | | | | | |
| | A | Autoregressive Integrated Moving Average (ARIMA) Models: Linear Models for Stationary Time Series, Stationary Time Series, Finite Order Moving Average (MA) Processes. | CO5 | | | | | | | | | | | |



| В | The First-Order Moving Average Process, MA(1), The Second- | CO5 |
|--------------|--|-----|
| | Order Moving Average Process, MA(2), Finite Order Autoregressive | |
| | Processes, First -Order Autoregressive Process, AR(l), Second-Order | |
| | Autoregressive Process, AR(2), | |
| C | General Autoregressive Process, AR(p), Partial Autocorrelation Function, PACF, Mixed Autoregressive-Moving Average CARMA) Processes, Time Series Model Building, Model Identification, Parameter Estimation, Examples of Building ARIMA Models, Forecasting ARIMA Processes. | CO5 |
| Unit 5 | | |
| А | Index Numbers: Definition, construction of index numbers, and problems thereof for weighted and unweighted index numbers including | CO6 |
| В | Laspeyre's, Paasche's, Edgeworth-Marshall, and Fisher's. Chain index numbers, | CO6 |
| С | Conversion of fixed-based to chain-based index numbers and vice- versa. Consumer price index numbers. | CO6 |
| Mode of | Theory | |
| examination | | |
| Weightage | CA 250/ | |
| Distribution | CA:25%; ESE:75% | |
| Text book/s* | 1. Gupta, S.C. and Kapoor, V.K., "Fundamental of Mathematical Statistics". | |
| Other | 1. Grewal, B.S, "Higher Engineering Mathematics". | |
| References | 2. Goon, A.M., Gupta, A.K. & Das Gupta. Fundamental of | |
| | Statistics. | |
| | | |

| РО | PO | PO | РО | PO | PO | PO | PO | PO | PO | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA110.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA110.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA110.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 3 | | 3 | |
| MDA110.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA110.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA110.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 3 | | 3 | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 3.0 | | 3.0 | |



| Scho | ool: SSBSR | Batch: 2024-28 | | |
|-----------------|--------------------------|--|--------------|--|
| | gramme: B.Sc. | Academic Year: 2027-28 | | |
| (Ho | | | | |
| | nch: Data Science | Semester: VII | | |
| α A 1 | nalytics Course Code | MDA111 | | |
| 2 | Course Title | | | |
| 2 | Credits | Non-Parametric Statistical Inference | | |
| | | 4 | | |
| 4 | Contact Hours (L-T-P) | 4-0-0 | | |
| | Course Status | DSE | | |
| 5 | Course | Familiarise students with basic concepts of non-parametric inferenc | e, | |
| | Objective | nonparametric estimation, order statistics use, and application in rea | l-life data. | |
| 6 | Course | CO1: Explain the concept of non-parametric inference. (K2, K4) | | |
| | Outcomes | CO2: Apply the concept of nonparametric estimation and explain th of the order statistic. (K3) | - | |
| | | CO3: Explain and use different non-parametric test estimators. (K2 CO4: Explain the properties of non-parametric test estimators.(K2, 1 | | |
| | | CO5: Describe the concept of order statistics. (K1, K2) CO6: Understand and evaluate the application of non-parametric | inference on | |
| 7 | | real-life data. (K2, K6) | · · · · · | |
| 7 | Course Description | This course will cover the basic concepts of non-parametron nonparametric estimation, order statistics use, and application in re- | | |
| 8 | | | | |
| | Unit 1 | | | |
| | А | Non-Parametric methods, Advantages and Disadvantages, | CO1 | |
| | В | Uses and application of the non-parametric method, | CO1 | |
| | С | Type of non-parametric test, | CO1 | |
| | Unit 2 | | | |
| | А | The sign test for paired data, One sample sign test, | CO2 | |
| | В | Ranked sum test, Mann-Whitney U test, | CO2 | |
| | С | Kruskalwali's test or H test, | CO2 | |
| | Unit 3 | | _ ~ _ | |
| | A | One sample run test, median test for randomness, | CO3 | |
| | В | Runs above and below the median, spearman rank correlation test | CO3, CO4 | |
| | С | Testing of hypothesis about rank correlation, | CO4 | |
| | Unit 4 | | | |
| | А | Kolmogrov Smirnov test, Kendall test of Concordance | CO5 | |
| | В | Median test for two independent samples, | CO5 | |
| | С | Wilcoxon Signed rank test, The Matched pairs sign, test | CO5 | |
| | Unit 5 | | | |
| | А | Introduction and application of order statistics, Distribution of Single Order Statistics, | CO6 | |
| | В | Joint distribution of two or more order statistics, Distribution of difference of two distinct order statistics. | CO6 | |
| | C | Distribution of Range, Distribution of Quartile, and Distribution of median. | CO6 | |
| | Mode of | Theory | | |
| | examination | | | |



| Weightage Distribution | CA:25%; ESE:75% |
|---------------------------|---|
| Text book/s* | 1. Gibbons, J.D. & Chakraborti, S.: Nonparametric Statistical Inference, 5th Edition. CRC Press. |
| | 2.Hollander, M., Wolfe, D. & Chicken, E. Nonparametric Statistical Methods, 3rd Edition. Wiley. |
| Other | 1.Bonnini, S., Corain, L., Marozzi, M. & Salmaso, L.: |
| References | Nonparametric Hypothesis Testing Rank and Permutation Methods with Applications in R. Wiley. |
| | 2. Sprent, P. & Smeeton, N.C. (2013): Applied |
| | Nonparametric Statistical Methods, 4th Edition. CRC |
| | Press. |

| РО | РО | PO | РО | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA111.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA111.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA111.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 3 | | 3 | |
| MDA111.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA111.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | | 3 | |
| MDA111.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 3 | | 3 | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 3.0 | | 3.0 | |



| Scho | ool: SSBSR | Batch: 2024-28 | |
|------------|-------------------------------|---|----------------------------------|
| | gramme: B.Sc. | Academic Year: 2027-28 | |
| (Ho | | | |
| | nch: Data Science nalytics | Semester: VII | |
| <u>a A</u> | Course Code | MDA112 | |
| 2 | Course Title | Econometrics | |
| 2 3 | Credits | 2 | |
| <u> </u> | Contact Hours | | |
| 4 | | 3-0-0 | |
| | (L-T-P) Course Status | DSE | |
| 5 | - | | |
| 3 | Course | The objective of this course is to introduce regression analysis to stude | ents so |
| (| Objective | that understand its applications in different fields of economics. | <u> </u> |
| 6 | Course Outcomes | CO1: Able to have concise knowledge of basic regression analysis or data and interpret and critically evaluate outcomes of empirical analys | |
| | Outcomes | K3). | IS. $(\mathbf{K}1, \mathbf{K}2,$ |
| | | CO2: Analyze the theoretical background for standard methods used i | n empirical |
| | | analyses, like properties of least squares estimators and statistical | |
| | | hypotheses. (K2, K3, K4). | |
| | | CO3: Able to apply for modern computer programs in regression | analyses of |
| | | empirical data, including statistical testing to investigate whether the | he classical |
| | | assumptions in regression analysis are satisfied. (K2, K3, K4). | |
| | | CO4: Design and development of a real-life model based on e | econometric |
| | | methods. (K4, K5, K6) | |
| | | CO5: Develop and apply advance methods for the implementation of e | |
| | | techniques also various functions for economic analysis and future (K5, K6). | forecasting. |
| | | CO6: Enable students to make use of econometric models in their acad | lemic work. |
| | | (K4,K5) | |
| 7 | Course | The purpose of this course is to give students a solid foundation in e | |
| | Description | techniques, various functions for economic analysis, and future forecas | |
| | | of the methods introduced in this course are also useful in business, f | inance, and |
| 8 | | many other disciplines. | |
| 0 | Unit 1 | | |
| | A | Introduction to econometrics. A review of least squares and | CO1 |
| | A | maximum likelihood estimation methods of parameters in the | COI |
| | | classical linear regression model and their properties. | |
| | | Generalized least squares estimation and prediction, construction of | CO1 |
| | В | confidence regions. | |
| | C | Tests of hypotheses, use of dummy variables, and seasonal | CO1 |
| | | adjustment. | |
| | Unit 2 | Degraggion analyzing under lingen northistigues upstricted linger | ~~~ |
| | A | Regression analysis under linear restrictions, restricted least squares estimation method and its properties. | CO2 |
| | В | Problem of Multicollinearity, its implications, and tools for | CO2 |
| | | handling the problem. | |
| | С | Ridge regression. Heteroscedasticity, consequences, and tests for it. | CO2 |
| | | | |
| | Unit 3 | | |
| | | Estimation procedures under heteroscedastic disturbances, Bartlett's test Breusch Pagan test, and Goldfelf Quandt test | CO3 |
| | Unit 3 A | Bartlett's test, Breusch Pagan test, and Goldfelf Quandt test. | |
| | Unit 3 | Estimation procedures under heteroscedastic disturbances, Bartlett's test, Breusch Pagan test, and Goldfelf Quandt test. Autocorrelation, sources, and consequences. Autoregressive process tests for autocorrelation. | CO3 CO3 CO4 |



| А | Durbin Watson test. Asymptotic theory and regressors. | CO5 | | | | | |
|--|---|-----|--|--|--|--|--|
| В | Instrumental variable estimation, errors in variables. | CO5 | | | | | |
| С | Simultaneous equations model, the problem of identification, a necessary and sufficient condition for the identifiability of parameters in a structural equation. | CO5 | | | | | |
| Unit 5 | | | | | | | |
| А | Ordinary least squares, indirect least squares. | CO6 | | | | | |
| В | Two-stage least square. | CO6 | | | | | |
| С | Limited information maximum likelihood method. | CO6 | | | | | |
| Mode of | Theory | | | | | | |
| examination | | | | | | | |
| Weightage | CA:25%; ESE:75% | | | | | | |
| Distribution | CA.2570, LSL.7570 | | | | | | |
| Text book/s* | 1.Gujrati, D.N. & Porter, D.C.: Basic Econometrics, 6th Edition. | | | | | | |
| | McGraw Hill. | | | | | | |
| | 2. Maddala, G.S. & Lahiri, K.: Introduction to Econometrics, 4th | | | | | | |
| | Edition. Wiley. | | | | | | |
| Other 1. Greene, W.H.: Econometric Analysis, 7th Edition. Pearson. | | | | | | | |
| References | 2. Studenmund, A.H. & Johnson, B.K.: Using Econometrics: A | | | | | | |
| | Practical Guide, 7th Edition. Pearson. | | | | | | |

| РО | РО | PO | PO | PO | PO | PO | PO | PO | РО | РО | PO | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|-----|----|----|-----|-----|-----|-----|
| СО | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA112.1 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 1 | 1 | 1 |
| MDA112.2 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 1 | 1 | 1 |
| MDA112.3 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 1 | 1 | 1 |
| MDA112.4 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 1 | 1 | 1 |
| MDA112.5 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 1 | 1 | 1 |
| MDA112.6 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 1 | 1 | 1 |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 3.0 | 1.0 | 1.0 | 1.0 |



| School: SSBSR | | Batch: 2024-28 | | | | | | | |
|---------------|--|--|--|--|--|--|--|--|--|
| (Hoi | | Academic Year: 2027-28 | | | | | | | |
| | nch: Data Science & lytics | | | | | | | | |
| 1 | Course Code | MDA113 | | | | | | | |
| 2 | Course Title | Survival Analysis | | | | | | | |
| 3 | Credits | 4 | | | | | | | |
| 4 | Contact Hours (L-T-P) | 4-0-0 | | | | | | | |
| | Course Status | DSE | | | | | | | |
| 5 | Course Objective | To demonstrate and intended to verse students in the techniques understand and carry out methods of research in survival analysis. | necessary to | | | | | | |
| 6 | Course Outcomes | CO1: Explain the concept of survival data, and the roles played by censoring, and survival and hazard functions. CO2: Format data appropriately for analysis, and understanding. CO3: Apply and drew the graph of survival data, and the Kaplan – Meier curve. CO4: Explain the concept ofKernel smoothed distribution estimator and kernel smoothed hazard rate estimator CO5: Describe how to fit the Cox Proportional Hazards model. CO6: Apply models to the data analysis using the Cox proportional hazards | | | | | | | |
| 7 | model.CourseA UG-level course in survival analysis, intended to verse stDescriptiontechniques necessary to understand and carry out methods of reseaanalysis. Lectures study the large-sample properties of estimatorssample, k-sample and partial likelihood inference, with proofs baseprocess and Martingale theory. The theory of competing risks isseveral angles. Many extensions of the Cox model to more | | | | | | | | |
| | | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is several angles. Many extensions of the Cox model to more | d on counting studied from | | | | | | |
| 8 | Outline syllabus | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is | d on counting studied from | | | | | | |
| 8 | Outline syllabus Unit 1 | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is several angles. Many extensions of the Cox model to more structures are considered. | d on counting studied from complex data CO Mapping | | | | | | |
| 8 | • | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is several angles. Many extensions of the Cox model to more structures are considered. Basic quantities. The survival functions. The hazard functions. The mean residual life time function and median life. | d on counting studied from complex data CO Mapping CO1 | | | | | | |
| 8 | Unit 1 A | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is several angles. Many extensions of the Cox model to more structures are considered. Basic quantities. The survival functions. The hazard functions. The mean residual life time function and median life. Common parametric models for survival data. Models for | d on counting studied from complex data CO Mapping CO1 | | | | | | |
| 8 | Unit 1 | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is several angles. Many extensions of the Cox model to more structures are considered. Basic quantities. The survival functions. The hazard functions. The mean residual life time function and median life. Common parametric models for survival data. Models for competing risks. Right censoring. Left or interval censoring. Truncation. Likelihood construction for censored and truncated data. Basic ideas for | d on counting studied from complex data CO Mapping CO1 CO1, CO2 | | | | | | |
| 8 | Unit 1 A B | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is several angles. Many extensions of the Cox model to more structures are considered. Basic quantities. The survival functions. The hazard functions. The mean residual life time function and median life. Common parametric models for survival data. Models for competing risks. Right censoring. Left or interval censoring. Truncation. Likelihood | d on counting studied from complex data CO Mapping CO1 CO1, CO2 | | | | | | |
| 8 | Unit 1 A B C | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is several angles. Many extensions of the Cox model to more structures are considered. Basic quantities. The survival functions. The hazard functions. The mean residual life time function and median life. Common parametric models for survival data. Models for competing risks. Right censoring. Left or interval censoring. Truncation. Likelihood construction for censored and truncated data. Basic ideas for counting processes and martingales. | d on counting studied from complex data CO Mapping CO1 CO1, CO2 CO1, CO2 | | | | | | |
| 8 | Unit 1 A B C Unit 2 | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is several angles. Many extensions of the Cox model to more structures are considered. Basic quantities. The survival functions. The hazard functions. The mean residual life time function and median life. Common parametric models for survival data. Models for competing risks. Right censoring. Left or interval censoring. Truncation. Likelihood construction for censored and truncated data. Basic ideas for counting processes and martingales. | d on counting studied from complex data CO Mapping CO1 CO1, CO2 CO1, CO2 CO3 CO3 | | | | | | |
| 8 | Unit 1 A B C Unit 2 A B C | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is several angles. Many extensions of the Cox model to more structures are considered. Basic quantities. The survival functions. The hazard functions. The mean residual life time function and median life. Common parametric models for survival data. Models for competing risks. Right censoring. Left or interval censoring. Truncation. Likelihood construction for censored and truncated data. Basic ideas for counting processes and martingales. Nonparametric estimators of the survival and cumulative hazard functions. Kaplan-Meier estimator and Nelson-Allen estimator. Point wise confidence intervals for the survival and cumulative | d on counting studied from complex data CO Mapping CO1 CO1, CO2 CO1, CO2 | | | | | | |
| 8 | Unit 1 A B C Unit 2 A B | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is several angles. Many extensions of the Cox model to more structures are considered. Basic quantities. The survival functions. The hazard functions. The mean residual life time function and median life. Common parametric models for survival data. Models for competing risks. Right censoring. Left or interval censoring. Truncation. Likelihood construction for censored and truncated data. Basic ideas for counting processes and martingales. Nonparametric estimators of the survival and cumulative hazard functions. Kaplan-Meier estimator and Nelson-Allen estimator. Point wise confidence intervals for the survival and cumulative hazard functions. Confidence bands for the survival function. Point and interval | d on counting studied from complex data CO Mapping CO1 CO1, CO2 CO1, CO2 CO3 CO3 | | | | | | |
| 8 | Unit 1 A B C Unit 2 A B C | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is several angles. Many extensions of the Cox model to more structures are considered. Basic quantities. The survival functions. The hazard functions. The mean residual life time function and median life. Common parametric models for survival data. Models for competing risks. Right censoring. Left or interval censoring. Truncation. Likelihood construction for censored and truncated data. Basic ideas for counting processes and martingales. Nonparametric estimators of the survival and cumulative hazard functions. Kaplan-Meier estimator and Nelson-Allen estimator. Point wise confidence intervals for the survival and cumulative hazard functions. Confidence bands for the survival function. Point and interval | d on counting studied from complex data CO Mapping CO1 CO1, CO2 CO1, CO2 CO3 CO3 | | | | | | |
| 8 | Unit 1 A B C Unit 2 A B C Unit 3 | sample, k-sample and partial likelihood inference, with proofs base process and Martingale theory. The theory of competing risks is several angles. Many extensions of the Cox model to more structures are considered. Basic quantities. The survival functions. The hazard functions. The mean residual life time function and median life. Common parametric models for survival data. Models for competing risks. Right censoring. Left or interval censoring. Truncation. Likelihood construction for censored and truncated data. Basic ideas for counting processes and martingales. Nonparametric estimators of the survival and cumulative hazard functions. Kaplan-Meier estimator and Nelson-Allen estimator. Point wise confidence intervals for the survival and cumulative hazard functions. Confidence bands for the survival function. Point and interval estimates of the mean and median survival time, and quintiles. | d on counting studied from complex data CO Mapping CO1 CO1, CO2 CO1, CO2 CO3 CO3 CO3 | | | | | | |



| Unit 4 | | |
|---------------------------|---|----------|
| А | Kernel smoothed distribution estimator and kernel smoothed hazard rate estimator. | CO4 |
| В | Hypothesis testing. One-sample tests. Tests for two samples and more than two samples. Tests for trend. Stratified log-rank test. | CO4 |
| С | Parametric models with covariates. The accelerated failure time (AFT) model. Some popular AFT models. Diagnostic methods for parametric models. | CO4 |
| Unit 5 | | |
| А | The Cox proportional hazards model. Partial likelihoods for distinct-event time data. | CO5, CO6 |
| В | Partial likelihood when ties are present. Local tests. Estimation of the survival function. | CO5, CO6 |
| С | Additional materials: Model building and high-dimensional data analysis using the Cox proportional hazards model. | CO5, CO6 |
| Mode of examination | Theory | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | 1.Lee, E. T. and Wang, J. W.: Statistical Methods for Survival Data Analysis, 3rdEdition. John Wiley. 2.Liu, X: Survival Analysis: Models and Applications, Wiley, New York. | |
| Other References | Kleinbaum, D. G. and Klein, M.: Survival Analysis: A Self- Learning Text, 3rdEd, Springer, New York. Hosmer, D. and Lemeshow, S.: Applied Survival Analysis: Regression Modeling of Time to Event Data, Wiley, New York. | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | PO | РО | PO | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|-----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA113.1 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA113.2 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA113.3 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA113.4 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA113.5 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA113.6 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 3.0 | | 1.0 | 1.0 |



| Scho | ol: SSBSR | Batch: 2024-28 | |
|------|--------------------------|---|--|
| | ramme: B.Sc. | Academic Year: 2027-28 | |
| (Hon | IS.) | | |
| | ich: Data Science | Semester: VII | |
| | nalytics | ND 4177 | |
| 1 | Course Code | MDA155 | |
| 2 | Course Title | Time Series, Forecasting and Index Number Lab | |
| 3 | Credits | 1 | |
| 4 | Contact Hours (L-T-P) | 0-0-2 | |
| | Course Status | DSE | |
| 5 | Course Objective | To provide students with hands-on experience in working with tim This includes exploring different types of time series data, unders characteristics, and learning how to preprocess and clean the data for 2. To familiarize the students with visualizing time series data u techniques such as line plots, scatter plots, seasonal decomposition plots. To help students gain insights into the patterns, trends, a variations present in the data. To familiarize the students with different time series techniques, such as autoregressive integrated moving averag models, exponential smoothing models, or state space models. The aim is to equip students with the knowledge and skills t apply appropriate models to analyze and forecast time series d | tanding their or analysis. sing various plots, and nd seasonal modelling e (ARIMA) o select and |
| 6 | Course Outcomes | The student will be able to select and apply appropriate models to forecast time series data. CO1: To familiarize the students to enter time series data in Excel/R a data transformation and adjustments. (K1, K2, K3) CO2: To find basic descriptive of the data and determining the trend time series methods. (K1, K2, K3) CO3: To find the least square estimates of the linear regression mode enable the students to check the model's adequacy. (K2, K3) CO4:To find the seasonal and cyclic variations in time series data.(K3, CO5: to predict new observations by applying ARIMA model (K4, K5) CO6: To enable students in employing Partial autocorrelation function auto-regressive moving average processes. (K4, K5, K6) | and do some I by various del and also , K4, K5) , K6) |
| 7 | Course Description | This is an advances course in statistics. Students are introduced to the involved in using sample data to make inferences about populations. I the study of measures of central tendency and dispersion, finite statistical inferences from large and small samples, linear regre correlation and hypothesis. | ncluded are probability, ession, and |
| 8 | Outline syllabus | | CO Mapping |
| | Unit 1 | Lab. Experiment 1 | 5 |
| | A, B, C | Problem-based how to enter time series data in a column, with each observation in a separate cell. Ensure the data is sorted in chronological order. Data transformation and adjustments. | CO1 |
| | Unit 2 | Lab. Experiment 2 | |
| | A, B, C | Problem-based on how to calculate basic descriptive statistics such as mean, median, and standard deviation. Analyze the data's trend by the method of the freehand curve, Moving average curve, semi-average | |



| | curve, and least square method. | | | | | | | | | | |
|---------------------------|--|----------|--|--|--|--|--|--|--|--|--|
| Unit 3 | Lab. Experiment 3 | | | | | | | | | | |
| A, B, C | Problem-based on Least square estimation in the linear regression model.Model Adequacy checking. Regression models for general time series data. Prediction of new observations in time series data. | CO3 | | | | | | | | | |
| Unit 4 | Lab. Experiment 4 | | | | | | | | | | |
| A, B, C | Problem-based on how to d etermine if data exhibits seasonality by calculating the seasonal indices. Methods for measuring linear trend Methods for measuring seasonal variations. Methods for measuring cyclic variations. | CO4 | | | | | | | | | |
| Unit 5 | Lab. Experiment 5 | | | | | | | | | | |
| A, B, C | Problem-based on how to use software to built-in forecasting functions to generate predictions. Linear models for stationary time series. Calculations of moving averages (first and second order). General auto-regressive process. Partial autocorrelation function. Mixed auto-regressive moving average processes. | CO5, CO6 | | | | | | | | | |
| Mode of examination | Practical+Viva | | | | | | | | | | |
| Weightage Distribution | CA:25%; CE:25%; ESE:50% | | | | | | | | | | |
| Text book/s* | Hyndman, R. J., & Athanasopoulos, G. (2018). Forecasting: principles and practice. Fuller, W. A. (2009). Introduction to statistical time series. John Wiley & Sons. | | | | | | | | | | |
| Other References | 1.Dan L. Shunk: Time Series Modeling for Analysis and Control: Advanced Autoregressive Techniques" | | | | | | | | | | |

| РО | РО | PO | PO | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|----|----|----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA155.1 | | | | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| MDA155.2 | | | | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| MDA155.3 | | | | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| MDA155.4 | | | | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| MDA155.5 | | | | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| MDA155.6 | | | | 2 | 1 | 1 | 1 | 3 | 1 | 1 | 3 | 1 | 2 | 3 |
| Average | | | | 2.0 | 1.0 | 1.0 | 1.0 | 3.0 | 1.0 | 1.0 | 3.0 | 1.0 | 2.0 | 3.0 |





| | ool: SSBSR | Batch: 2024-28 | | | | | | | | | | | |
|-----|-------------------------------|---|--|--|--|--|--|--|--|--|--|--|--|
| (Ho | | Academic Year: 2027-28 | | | | | | | | | | | |
| | nch: Data Science & lytics | | | | | | | | | | | | |
| 1 | Course Code | Econometrics Lab | | | | | | | | | | | |
| 2 | Course Title | MDA156 | | | | | | | | | | | |
| 3 | Credits | 1 | | | | | | | | | | | |
| 4 | Contact Hours(L- T-P) | 0-0-2 | | | | | | | | | | | |
| | Course Status | DSE | | | | | | | | | | | |
| 5 | Course Objective | To enable the student in understanding and apply mathematical techniques to economic data in R/Excel To enable students to identify the causal relationship and magnitude of these relationships. To make Students learn how to specify appropriate economet capture the relationships between economic variables To enable Students how to collect, clean, and preprocess exploratory data analysis, and apply econometric techniques to interpret the results. To familiarize the students to assess the statistical signal relationships and variables using Hypothesis testing. | quantify the ric models to data, conduct estimate and | | | | | | | | | | |
| 6 | Course Outcomes | The student will be able to do exploratory data analysis of a time see CO1: to find the estimates of the parameters using least square e maximum likelihood estimates. (K1, K2, K3) CO2: to find the confidence interval and test for significance of t of the parameters of classical linear regression. (K1, K2, K3) CO3: to solve the Linear non-homogeneous PDE with constant (K2, K3) CO4: to employ Regression analysis under linear restriction and of for Multicollinearity. (K3, K4, K5) CO5: to check whether data is having Heteroscedasticity by appli- methods. (K4, K5, K6) CO6: to determine whether there is autocorrelation in the data by u tests. (K4, K5, K6) | estimates and he estimates t coefficient. employ tests ying various using various | | | | | | | | | | |
| 7 | Course | The course is an introduction to R/Excel in Econometrics. 7 | | | | | | | | | | | |
| | Description | objective of the course is to develop basic knowledge of employing | ng statistical | | | | | | | | | | |
| | | techniques to economic data | | | | | | | | | | | |
| 8 | Outline syllabus | | CO Mapping | | | | | | | | | | |
| | Unit 1 | Lab. Experiment 1 | | | | | | | | | | | |
| | A, B, C | Problem-based on estimation of parameters of classical linear regression by maximum likelihood estimation(MLEs), Least square estimation(LSE), Generalized least square estimation | CO1, CO2 | | | | | | | | | | |
| | Unit 2 | Lab. Experiment 2 | | | | | | | | | | | |
| | A, B, C | Problem-based on Confidence interval of parameters, Test for the significance of estimates of the parameters. Use of dummy variable and seasonal adjustment. | CO2, CO3 | | | | | | | | | | |
| | Unit 3 | Lab. Experiment 3 | | | | | | | | | | | |
| | A, B, C | Problem-based on Regression analysis under linear restriction Restricted least square estimation. Multicollinearity: test and tools to handle this problem | CO3, CO4 | | | | | | | | | | |
| | Unit 4 | Lab. Experiment 4 | | | | | | | | | | | |
| | A, B, C | Problem-based on Heteroscedastic disturbances tests; Bartlett's test, Breusch pagan Test, Goldfelf Quandt test. | CO5, CO6 | | | | | | | | | | |



| I | Unit 5 | Lab. Experiment 5 | |
|-----|---------------------------|---|----------|
| A | A, B, C | Problem-based Autocorrelation sources; Autoregressive tests for autocorrelation. Durbin Watson test, Ordinary least square, indirect least square. | CO5, CO6 |
| | Mode of examination | Practical + Viva | |
| | Weightage Distribution | CA:25%; CE:25%; ESE:50% | |
|]] | Γext book/s* | Gujrati, D.N. & Porter, D.C.: Basic Econometrics, 6th Edition. McGraw Hill. Maddala, G.S. & Lahiri, K.: Introduction to Econometrics, 4th Edition. Wiley. | |
| | Other References | Greene, W.H.: Econometric Analysis, 7th Edition. Pearson. Studenmund, A.H. &Johnson, B.K.: Using Econometrics: A Practical Guide, 7th Edition. Pearson. | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | РО | РО | PO | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|-----|-----|-----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA156.1 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| MDA156.2 | 1 | 2 | 3 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| MDA156.3 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| MDA156.4 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| MDA156.5 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| MDA156.6 | 1 | 2 | 2 | 2 | | 1 | 1 | 3 | 1 | | 3 | 1 | 2 | 3 |
| Average | 1.0 | 2.0 | 2.0 | 2.0 | | 1.0 | 1.0 | 3.0 | 1.0 | | 3.0 | 1.0 | 2.0 | 3.0 |



| Scho | ol: SSBSR | Batch: 2024-28 | | | |
|-------------|-------------------------------|--|------------------|--|--|
| | gramme: B.Sc. | Academic Year: 2027-28 | | | |
| (Hon | | | | | |
| | ich: Data Science nalytics | Semester: VIII | | | |
| a Al | Course Code | MDA107 | | | |
| 2 | Course Title | Advanced Big Data and Text Analytics | | | |
| 3 | Credits | A A A A A A A A A A A A A A A A A A A | | | |
| 4 | Contact Hours | | | | |
| 4 | (L-T-P) | 4-0-0 | | | |
| | Course Status | CC | | | |
| 5 | Course | This course aims to provide insight into the concepts of Nati | iral Language | | |
| 5 | Objective | Processing and its applications. This course helps the students to in | | | |
| | Objective | applications using deep learning algorithms. This course helps | | | |
| | | various word/text representation algorithms. | | | |
| 6 | Course | At the end of the course, the student should be able to | | | |
| | Outcomes | CO1: Learn about Big data techniques and their applications. | | | |
| | | CO2: Analyse various neural network problems. | | | |
| | | CO3: Use different word/text representation methods to see how wo to each other. | ords are related | | |
| | | CO4: Model different NLP applications using Machine Learning/ | Deen learning | | |
| | | algorithms | Deep learning | | |
| | | CO5: Implement different deep learning models to solve real-time NI | LP problems | | |
| | | CO6: Provide a body of concepts and techniques for designing intelli | | | |
| 7 | Course | A UG-level course in Soft Computing Techniques to Improve Big | Data Analysis | | |
| | Description | solutions is to strengthen the dialogue between the statistics and s | | | |
| | • | research communities. | 1 0 | | |
| 8 | Outline syllabus | | CO Mapping | | |
| | Unit 1 | | mapping | | |
| | A | Introduction to Big Data: Introduction to Big Data, Big Data | CO1 | | |
| | | characteristics | | | |
| | | Types of Big Data, Structured Data, Unstructured Data, and semi | CO1 | | |
| | В | Structured Data. | | | |
| | С | Traditional vs. Big Data business approach, Case Study of Big Data | CO1 | | |
| | | Solutions. | | | |
| | Unit 2 | | | | |
| | А | Mining Data Streams: The Stream Data Model: A Data Stream- | CO2 | | |
| | | Management System, Examples of Stream Sources, Stream Queries, | | | |
| | | Issues in Stream Processing. | | | |
| | В | Sampling Data in a Stream: Obtaining a Representative Sample, The | CO2 | | |
| | | General Sampling Problem, Varying the Sample Size. Filtering | | | |
| | С | Streams: The Bloom Filter, Analysis. Counting Distinct Elements in a Stream The Count-Distinct Problem. | CO2 | | |
| | C | The Flajolet-Martin Algorithm, Combining Estimates, Space | ,002 | | |
| | | Requirements Counting Ones in a Window: The Cost of Exact | | | |
| | | Counts. | | | |
| | Unit 3 | | | | |
| | А | The Big Data Analytics and Big Data Analytics Techniques: Big | CO3 | | |
| | | Data and its Importance, Drivers for Big data, Optimization | | | |
| | | techniques, Dimensionality Reduction techniques. | | | |
| | | | | | |
| | В | Time series Forecasting, Social Media Mining, and Social Network Analysis, and its Application. | CO3 | | |



| С | Big Data analysis using Hadoop, Pig, Hive, MongoDB, Spark, and Mahout, Data analysis techniques. | CO3 |
|---------------------------|---|-----|
| Unit 4 | | |
| A | Introduction to Natural Language Processing Words Regular Expressions N-grams Language modeling Part of Speech. | CO4 |
| В | Tagging Named Entity Recognition Syntactic and Semantic Parsing- Morphological Analysis | CO4 |
| С | Text Representation and Transformation-Vector space models Bag of Words Term Frequency Inverse Document Frequency Word Vector representations: Word2vec, GloVe, FastText, BERT-Topic Modelling | CO4 |
| Unit 5 | | |
| А | Neural language models - Recurrent Neural Network - Long Short- Term Memory Networks | CO5 |
| В | Encoder decoder architecture - Attention Mechanism - Transformer networks | CO6 |
| С | Text classification-Sentiment Analysis-Neural Machine Translation - Question answering - Text summarization | CO6 |
| Mode of examination | Theory | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | S.N. Sivanandam& S.N. Deepa, Principles of Soft Computing, Wiley Publications, 2nd Edition. S. Rajasekaran& G.A. VijayalakshmiPai, Neural Networks, | |
| Other References | N. K. Bose, Ping Liang, Neural Network fundamental with Graph, Algorithms & Applications, TMH, 1st Edition. Martin T Hagen, Neural Network Design, Nelson Candad, 2nd Edition. | |

| РО | РО | PO | РО | PO | PO | PO | PO | PO | PO | РО | РО | PSO | PSO | PSO |
|----------|-----|-----|-----|-----|----|-----|----|----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA107.1 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | | | |
| MDA107.2 | 2 | 3 | 3 | 2 | | 1 | | | | | 3 | | | |
| MDA107.3 | 2 | 2 | 2 | 3 | | 1 | | | | | 3 | | | |
| MDA107.4 | 2 | 3 | 2 | 2 | | 1 | | | | | 3 | | | |
| MDA107.5 | 3 | 3 | 2 | 2 | | 1 | | | | | 3 | | | |
| MDA107.6 | 3 | 3 | 2 | 3 | | 1 | | | | | 3 | | | |
| Average | 2.3 | 2.6 | 2.0 | 2.1 | | 1.0 | | | | | 3.0 | | | |



| Sch | ool: SSBSR | Batch: 2024-28 | | | | | | | | | | |
|-----|--------------------------------|--|---------------|--|--|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2027-28 | | | | | | | | | | |
| (Ho | ns.) nch: Data Science | | | | | | | | | | | |
| | nch: Data Science Analytics | Semester: VIII | | | | | | | | | | |
| 1 | Course Code | MDA114 | | | | | | | | | | |
| 2 | Course Title | Bayesian Data Analysis | | | | | | | | | | |
| 3 | Credits | 4 | | | | | | | | | | |
| 4 | Contact Hours (L-T-P) | 4-0-0 | | | | | | | | | | |
| | Course Status | CC | | | | | | | | | | |
| 5 | Course Objective | To make students familiar with the concepts of preparing Working with dates and times, Data Cleaning, Data Structure, an Text Data. | • | | | | | | | | | |
| 6 | Course | CO1: Explain in detail the Bayesian framework for data ana | lysis and its | | | | | | | | | |
| | Outcomes | flexibility and be able to demonstrate when the Bayesian appr | oach can be | | | | | | | | | |
| | | beneficial. | | | | | | | | | | |
| | | CO2: Develop, analytically describe, and implement both singl | e and Multi- | | | | | | | | | |
| | | Parameter probability models in the Bayesian framework. | | | | | | | | | | |
| | | CO3: Demonstrate the role of the prior distribution in Bayesian inference | | | | | | | | | | |
| | | and be able to articulate the usage of non-informative priors and conjugate | | | | | | | | | | |
| | | | | | | | | | | | | |
| | | priors. | | | | | | | | | | |
| | | CO4: Show high level Interpretation of Bayesian Analysis Re | sults and be | | | | | | | | | |
| | | able to readily perform Bayesian model evaluation and assessme | nt. | | | | | | | | | |
| | | 205: Demonstrate the necessary skills to: fit hierarchical models, provide | | | | | | | | | | |
| | | thorough technical specifications for these models. | | | | | | | | | | |
| | | CO6: Bayesian statistical practice makes extensive use of versions of | | | | | | | | | | |
| | | objective b.ayesian analysis | | | | | | | | | | |
| 7 | Course Description | This course introduces preparing your data; Working with dates Data Cleaning, Data Structure, and cleaning Text Data. | and times, | | | | | | | | | |
| 8 | | - | | | | | | | | | | |
| | Unit 1 | | | | | | | | | | | |
| | A | Limitations of empirical and logical theories of probability | CO1 | | | | | | | | | |
| | В | Subjective probability, determination of subjective probability, likelihood function, prior distribution, posterior distribution | CO1 | | | | | | | | | |
| | С | Bayes' theorem, methods of construction of priors and computation of the posterior distribution. | CO1 | | | | | | | | | |
| | Unit 2 | | | | | | | | | | | |
| | A | Natural conjugate family of priors for a model. | CO2 CO2 | | | | | | | | | |
| | B C | Hyper parameters of a prior from conjugate family. Conjugate families for (i) exponential family models, (ii) models admitting sufficient statistics of fixed dimension. | CO2 | | | | | | | | | |
| | Unit 3 | inders withting sufficient statistics of fixed differsion. | | | | | | | | | | |
| | | | | | | | | | | | | |



| r | | | |
|---|--------------|---|------|
| | А | Enlarging the natural conjugate family by (i) enlarging hyper parameter space (ii) mixtures from conjugate family | CO3 |
| | В | Choosing an appropriate member of conjugate prior family. | CO3 |
| | С | Non-informative, improper and invariant priors. Jeffrey's invariant prior. | CO3C |
| | Unit 4 | | |
| | А | Bayesian point estimation: As a prediction problem from posterior distribution. | CO4 |
| | В | Bayes estimators for (i) absolute error loss (ii) squared error loss (iii) 0-1 loss function. | CO4 |
| | С | Generalization to convex loss functions. Evaluation of the estimate in terms of the posterior risk. | CO4 |
| | Unit 5 | | |
| | А | Bayesian interval estimation: Credible intervals. | CO5 |
| | В | Highest posterior density regions. Interpretation of the | |
| | | confidence coefficient of an interval and its comparison with the interpretation of the confidence coefficient for a classical confidence interval. | CO5 |
| | С | Bayesian testing of hypothesis: Specification of appropriate | |
| | C | form of the prior distribution for a Bayesian testing of hypothesis problem. | CO6 |
| | Mode of | Theory | |
| | examination | Тпеогу | |
| | Weightage | | |
| | Distribution | CA:25%; ESE:75% | |
| | Text book/s* | 1. Bad Data Handbook: Cleaning Up the Data So You Can Get | |
| | TCAT UOUK/S | Back to Work by Q. Ethan McCallum 2. Best Practices in Data Cleaning: A Complete Guide to Everything You Need to Do Before and After Collecting Your Data by Jason W Osborne | |
| | Other | 1. Data Wrangling with Python by Jacqueline Kazil | |
| | References | 2. Principles of Data Wrangling: Practical Techniques for Data Preparation by Tye Rattenbury | |

| РО | РО | PO | PO | PO | PO | PO | PO | PO | РО | РО | РО | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|-----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA114.1 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA114.2 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA114.3 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA114.4 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA114.5 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA114.6 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 3.0 | | 1.0 | 1.0 |



| Scł | 1001: SSBSR | Batch: 2024-28 | | | | | | | | |
|-----|---|---|--|--|--|--|--|--|--|--|
| | ogramme: B.Sc. | Academic Year: 2027-28 | | | | | | | | |
| | ons.) anch: Data Science | a Compostore VIII | | | | | | | | |
| | Analytics | eSemester: vIII | | | | | | | | |
| 1 | Course Code | MDA117 | | | | | | | | |
| 2 | Course Title | Computational Intelligence | | | | | | | | |
| 3 | Credits | 4 | | | | | | | | |
| 4 | Contact Hours | 4-0-0 | | | | | | | | |
| I | (L-T-P) | | | | | | | | | |
| | Course Status | СС | | | | | | | | |
| 5 | Course | To provide a strong foundation on fundamental concepts in Com | outational | | | | | | | |
| 0 | Objective | Intelligence. | , automai | | | | | | | |
| | objective | To enable Problem-solving through various searching techniques | | | | | | | | |
| | | | | | | | | | | |
| 6 | Course | CO1: Provide a basic exposition to the goals and me | thods of | | | | | | | |
| | Outcomes | Computational Intelligence. | | | | | | | | |
| | | CO2: Study of the design of intelligent computational techniques. | | | | | | | | |
| | | CO3: Apply the Intelligent techniques for problem solving | | | | | | | | |
| | | CO4: Improve problem solving skills using the acquired knowled | | | | | | | | |
| | | areas of reasoning, natural language understanding, compute | er vision, | | | | | | | |
| | | automatic programming and machine learning. | | | | | | | | |
| | | CO5: Learn about the advance concept of AI | 1 1 | | | | | | | |
| | | CO6: Explain computable functions, predicates, forward and backward reasoning | | | | | | | | |
| 7 | Course | To apply these techniques in applications which involve po | ercention | | | | | | | |
| / | Description | reasoning and learning. To apply Computational Intelligence to | | | | | | | | |
| | 200000 | for information retrieval. To apply Computational Intelligence to | | | | | | | | |
| | | primarily for machine learning. | connques | | | | | | | |
| 8 | Outline syllabus | | CO | | | | | | | |
| | | | Mappi | | | | | | | |
| | | | ng | | | | | | | |
| | Unit 1 | | | | | | | | | |
| | А | Introduction to Artificial Intelligence-Search-Heuristic | | | | | | | | |
| | | e | CO1, | | | | | | | |
| | | Search A* algorithm Game Playing Alpha Beta Pruning Expert | - | | | | | | | |
| | В | = | - | | | | | | | |
| | B C | Search A* algorithm Game Playing Alpha Beta Pruning Expert | CO1, | | | | | | | |
| | С | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems | CO1, | | | | | | | |
| | | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems Inference Rules Forward Chaining and Backward Chaining Genetic Algorithms | CO1, CO1, | | | | | | | |
| | С | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems Inference Rules Forward Chaining and Backward Chaining Genetic Algorithms Proposition Logic First Order Predicate Logic Unification | CO1, CO1, | | | | | | | |
| | C Unit 2 A | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems Inference Rules Forward Chaining and Backward Chaining Genetic Algorithms Proposition Logic First Order Predicate Logic Unification Forward Chaining | CO1, ;CO1, ;CO2 | | | | | | | |
| | C Unit 2 | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems Inference Rules Forward Chaining and Backward Chaining Genetic Algorithms Proposition Logic First Order Predicate Logic Unification Forward Chaining Backward Chaining Resolution Knowledge Representation | CO1, ;CO1, ;CO2 | | | | | | | |
| | C Unit 2 A B | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems Inference Rules Forward Chaining and Backward Chaining Genetic Algorithms Proposition Logic First Order Predicate Logic Unification Forward Chaining Backward Chaining Resolution Knowledge Representation Ontological Engineering Categories and Objects | CO1, CO1, CO2 CO2 | | | | | | | |
| | C Unit 2 A | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems Inference Rules Forward Chaining and Backward Chaining Genetic Algorithms Proposition Logic First Order Predicate Logic Unification Forward Chaining Backward Chaining Resolution Knowledge Representation Ontological Engineering Categories and Objects Event Mental Events and Mental Objects Reasoning Systems for | CO1, CO1, CO2 CO2 CO3 | | | | | | | |
| | C Unit 2 A B | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems Inference Rules Forward Chaining and Backward Chaining Genetic Algorithms Proposition Logic First Order Predicate Logic Unification Forward Chaining Backward Chaining Resolution Knowledge Representation Ontological Engineering Categories and Objects Event Mental Events and Mental Objects Reasoning Systems for Categories Reasoning with Default Information Prolog | CO1, CO1, CO2 CO2 CO3 | | | | | | | |
| | C Unit 2 A B C | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems Inference Rules Forward Chaining and Backward Chaining Genetic Algorithms Proposition Logic First Order Predicate Logic Unification Forward Chaining Backward Chaining Resolution Knowledge Representation Ontological Engineering Categories and Objects Event Mental Events and Mental Objects Reasoning Systems for | CO1, CO1, CO2 CO2 CO3 | | | | | | | |
| | C Unit 2 A B C Unit 3 | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems Inference Rules Forward Chaining and Backward Chaining Genetic Algorithms Proposition Logic First Order Predicate Logic Unification Forward Chaining Backward Chaining Resolution Knowledge Representation Ontological Engineering Categories and Objects Event Mental Events and Mental Objects Reasoning Systems for Categories Reasoning with Default Information Prolog Programming. | CO1, CO1, CO2 CO2 CO3 | | | | | | | |
| | C Unit 2 A B C | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems Inference Rules Forward Chaining and Backward Chaining Genetic Algorithms Proposition Logic First Order Predicate Logic Unification Forward Chaining Backward Chaining Resolution Knowledge Representation Ontological Engineering Categories and Objects Event Mental Events and Mental Objects Reasoning Systems for Categories Reasoning with Default Information Prolog | CO1, CO1, CO2 CO2 CO3 | | | | | | | |
| | C Unit 2 A B C Unit 3 | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems Inference Rules Forward Chaining and Backward Chaining Genetic Algorithms Proposition Logic First Order Predicate Logic Unification Forward Chaining Backward Chaining Resolution Knowledge Representation Ontological Engineering Categories and Objects Event Mental Events and Mental Objects Reasoning Systems for Categories Reasoning with Default Information Prolog Programming. | CO1, CO1, CO2 CO2 CO3 | | | | | | | |
| | C Unit 2 A B C Unit 3 A | Search A* algorithm Game Playing Alpha Beta Pruning Expert systems Inference Rules Forward Chaining and Backward Chaining Genetic Algorithms Proposition Logic First Order Predicate Logic Unification Forward Chaining Backward Chaining Resolution Knowledge Representation Ontological Engineering Categories and Objects Event Mental Events and Mental Objects Reasoning Systems for Categories Reasoning with Default Information Prolog Programming. | CO1, CO1, CO2 CO2 CO3 CO4 | | | | | | | |



| Unit 4 | |
|--------------|---|
| А | Probability basics - Bayes Rule and its Applications BayesianCO5 |
| | Networks Exact and Approximate Inference in Bayesian |
| | Networks Hidden Markov Models Forms of Learning |
| В | Supervised Learning - Learning Decision Trees – Regression and CO5 |
| | Classification with Linear Models - Artificial Neural Networks – |
| C | Nonparametric Models Support Vector Machines Statistical CO5 |
| | Learning, Learning with Complete Data Learning with Hidden |
| | Variables- The EM Algorithm Reinforcement Learning. |
| Unit 5 | |
| А | Natural language processing-Morphological Analysis SyntaxCO6 analysis |
| В | Semantic Analysis All applications Language ModelsCO6 Information Retrieval Information |
| С | Extraction Machine Translation Machine Learning Symbol CO6 Based Machine Learning: Connectionist Machine Learning. |
| Mode of | Theory |
| examination | |
| Weightage | |
| Distribution | CA:25%; ESE:75% |
| Text book/s* | 1. Stuart Russell, Peter Norvig, Artificial Intelligence: A Modern |
| | Approach, Third Edition, Pearson Education / Prentice Hall of |
| | India. |
| | 2. Elaine Rich and Kevin Knight, Artificial Intelligencel, Third |
| | Edition, Tata McGraw- Hill. |
| Other | 1. Patrick H. Winston. "Artificial Intelligence", Third edition, |
| References | Pearson Edition. |
| | 2. Dan W. Patterson, Introduction to Artificial Intelligence and Expert Systems, PHI. |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | PO | РО | PO | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|-----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA117.1 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA117.2 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA117.3 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA117.4 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA117.5 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| MDA117.6 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | | 1 | 1 |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 3.0 | | 1.0 | 1.0 |



| School: SSBSR | | Batch: 2024-28 | | | | | | | | | |
|---------------|------------------------------|--|-----------------|--|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2027-28 | | | | | | | | | |
| (Ho | | | | | | | | | | | |
| | nch: Data Scienc nalytics | e Semester: VIII | | | | | | | | | |
| 1 1 | Course Code | MDA115 | | | | | | | | | |
| 2 | Course Title | Demography | | | | | | | | | |
| 3 | Credits | 4 | | | | | | | | | |
| 4 | Contact Hours | | | | | | | | | | |
| • | (L-T-P) | 4-0-0 | | | | | | | | | |
| | Course Status | DSE | | | | | | | | | |
| 5 | Course | The course tends to develop a basic understanding of demographic | c theory and | | | | | | | | |
| | Objective | its application to various aspects of the economy. The course will | | | | | | | | | |
| | 5 | presenting an economic argument and develop analytical abilities | of different | | | | | | | | |
| | | demographic concepts in quantitative terms. | | | | | | | | | |
| 6 | Course | CO1: Gain a sound command over the basic tenets of demograph | | | | | | | | | |
| | Outcomes | key demographic issues and illustrations in the context of a large | and diverse | | | | | | | | |
| | | country like India. | | | | | | | | | |
| | | CO2: Grasp a clear understanding of the inter-relationsh demography and the process of economic development. | ip between | | | | | | | | |
| | | CO3: Comprehend the basic components of population (fertility | . mortality | | | | | | | | |
| | | migration) | , 111011411117, | | | | | | | | |
| | | CO4: To study established theories of population. | | | | | | | | | |
| | | CO5: To explore various aspects of the population policy and | to study its | | | | | | | | |
| | | impact on socio economic issues. | | | | | | | | | |
| | | CO6: Identify appropriate sources of data, perform basic d | | | | | | | | | |
| | | analyses using various techniques and ensure their comparab | oility across | | | | | | | | |
| 7 | Course | populations. This course provides an introduction to demography and population | n studies | | | | | | | | |
| / | Description | This course provides an infoduction to demography and populate | ni studies | | | | | | | | |
| 8 | | | | | | | | | | | |
| | Unit 1 | Introduction | | | | | | | | | |
| | A | Demography- Its definition, nature and scope, its relation with | CO1 | | | | | | | | |
| | | other disciplines. | | | | | | | | | |
| | | Theories of population-Malthusian Theory, Optimum theory of | CO1 | | | | | | | | |
| | В | population and theory of Demographic Transition. | | | | | | | | | |
| | С | Population growth in India, Features of Indian Population. | CO1 | | | | | | | | |
| | | | 001 | | | | | | | | |
| | Unit 2 | Sources of Demographic data in India | | | | | | | | | |
| | А | Salient features of census- including 2011 census, Civil | CO2 | | | | | | | | |
| | | Registration System. | | | | | | | | | |
| | В | National Sample Survey | CO2 | | | | | | | | |
| | С | Demographic Survey- National Family Health Survey – 1, 2 and | CO2 | | | | | | | | |
| | | 3 Relative merits and demerits of these sources. | | | | | | | | | |
| | Unit 3 | Techniques of Analysis | | | | | | | | | |
| | A | Crude birth rate and death rate, Age specific birth rate and death | CO3 | | | | | | | | |
| | | rate, standardized birth rate and death rate. | | | | | | | | | |
| | В | Study of fertility- Total Fertility Rate, Gross Reproduction | CO3 | | | | | | | | |
| | | Rate and Net Reproduction Rate | | | | | | | | | |
| | С | Measurement of Population Growth rate- Simple Growth Rate | CO3 | | | | | | | | |
| | 1 | and Compound Growth Rate. | | | | | | | | | |



| Unit 4 | Modals of Demography& Life table | |
|---------------------------|--|-----|
| А | Logistic Models, Measures of Morbidity, Mortality graduation | CO4 |
| В | Methods of Construction of Abridged life Tables and its Applications. | CO4 |
| С | Population Estimates and Projection. | CO4 |
| Unit 5 | Vital Statistics | |
| A | Vital Statistics: Historical background, Civil Registration System in India: history, coverage, problems of civil registration, Sample Registration System (SRS), advantages and limitations. | CO5 |
| В | Population Surveys: Meaning, Scope, uses, limitations; Major surveys: National Sample Surveys (NSS), World Fertility Survey (WFS). | CO5 |
| С | Demographic Health Surveys (DHS), Reproductive and Child Health Survey (RCHS). National Family Health Surveys (NFHS), Comprehensive Nutrition Survey; Aging survey | CO6 |
| Mode of | Theory | |
| examination | | |
| Weightage Distribution | CA:25%; ESE:75% | |
| Text book/s* | Agarwal S.S.: India's Population Problem- Tata McGraw Hill Publication, Bombay Bhende A.A. and Tara Kanitkar: 'Principles of Population Studies'- Himalaya Publishing House, Bombay | |
| Other | 1.Hans Raj: 'Fundamentals of Demography'-Surjeet | |
| References | Publication, Delhi 2. Srinivasan K.: 'Basic Demographic Techniques and Applications', Sage Publications, New Delhi. | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | PO | РО | PO | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|-----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA115.1 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 3 | |
| MDA115.2 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 3 | |
| MDA115.3 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 3 | |
| MDA115.4 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 3 | |
| MDA115.5 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 3 | |
| MDA115.6 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 3 | |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 3.0 | 3.0 | 3.0 | |



| School: SSBSR | | Batch: 2024-28 | | | | | | | | |
|---------------|--|---|--|--|--|--|--|--|--|--|
| | gramme: B.Sc. | Academic Year: 2027-28 | | | | | | | | |
| (H0 Pro | ons.) Inch: Data Science & | - Somostovi VIII | | | | | | | | |
| | alytics | z semester: vill | | | | | | | | |
| 1 | Course Code | MDA116 | | | | | | | | |
| 2 | Course Title | Statistical Quality Control | | | | | | | | |
| 3 | Credits | 4 | | | | | | | | |
| 4 | Contact Hours(L- | • | | | | | | | | |
| 7 | T-P) | 4-0-0 | | | | | | | | |
| | Course Status | DSE | | | | | | | | |
| 5 | Course | The course tends to a comprehensive coverage of modern qu | ality control | | | | | | | |
| 5 | | techniques to include the design of statistical process control | | | | | | | | |
| | Objective | Objective techniques to include the design of statistical process co- acceptance sampling, and process improvement. | | | | | | | | |
| 6 | Course | CO1: Acquire knowledge and develop analysis skills of | on industria | | | | | | | |
| U | Outcomes | experimentation. | in maastra | | | | | | | |
| | | CO2: Acquire knowledge on acceptance sampling principles a | nd methods. | | | | | | | |
| | | CO3: Develop skills to analyse quality related data usin | | | | | | | | |
| | | statistical methods. | e | | | | | | | |
| | | CO4: Acquire knowledge on the traditional statistical qu | ality control | | | | | | | |
| | | methods and develop charting techniques. | | | | | | | | |
| | | CO5: Become familiar with the advanced statistical qua | ality control | | | | | | | |
| | | methods. | | | | | | | | |
| | | CO6: Develop new empirical approaches to quality related problems. | | | | | | | | |
| 7 | Course | This course provides an introduction to Statistical Quality Cor | ntrol. | | | | | | | |
| | Description | | | | | | | | | |
| 8 | | | | | | | | | | |
| | Unit 1 | Introduction of Quality Control | | | | | | | | |
| | А | Quality: Definition Its concept, application and importance. | CO1 | | | | | | | |
| | | Introduction to Process and Product Controls. | | | | | | | | |
| | | Seven tools of SDC, shares and assignable Causes of quality | 001 | | | | | | | |
| | | - Seven loois of SPU, chance and assignable Causes of quality | COL | | | | | | | |
| | В | Seven tools of SPC, chance and assignable Causes of quality variation. Statistical Control Charts. | CO1 | | | | | | | |
| | В | variation. Statistical Control Charts. | | | | | | | | |
| | B C | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, | C01 | | | | | | | |
| | C | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. | | | | | | | | |
| | C Unit 2 | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts | CO1 | | | | | | | |
| | C | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s- | | | | | | | | |
| | C Unit 2 A | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. | CO1 CO2 | | | | | | | |
| | C Unit 2 | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and | CO1 | | | | | | | |
| | C Unit 2 A B | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and u-chart. | CO1 CO2 CO2 | | | | | | | |
| | C Unit 2 A | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and u-chart. Comparison between control charts for variables and control | CO1 CO2 | | | | | | | |
| | C Unit 2 A B | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and u-chart. Comparison between control charts for variables and control charts for attributes. Analysis of patterns on control chart, | CO1 CO2 CO2 | | | | | | | |
| | C Unit 2 A B C | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and u-chart. Comparison between control charts for variables and control charts for attributes. Analysis of patterns on control chart, estimation of process capability. | CO1 CO2 CO2 | | | | | | | |
| | C Unit 2 A B C Unit 3 | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and u-chart. Comparison between control charts for variables and control charts for attributes. Analysis of patterns on control chart, estimation of process capability. Techniques of Analysis | CO1 CO2 CO2 CO2 | | | | | | | |
| | C Unit 2 A B C | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and u-chart. Comparison between control charts for variables and control charts for attributes. Analysis of patterns on control chart, estimation of process capability. Techniques of Analysis Crude birth rate and death rate, Age specific birth rate and | CO1 CO2 CO2 | | | | | | | |
| | C Unit 2 A B C Unit 3 A | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and u-chart. Comparison between control charts for variables and control charts for attributes. Analysis of patterns on control chart, estimation of process capability. Techniques of Analysis Crude birth rate and death rate, Age specific birth rate and death rate. | CO1 CO2 CO2 CO2 CO3 | | | | | | | |
| | C Unit 2 A B C Unit 3 | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and u-chart. Comparison between control charts for variables and control charts for attributes. Analysis of patterns on control chart, estimation of process capability. Techniques of Analysis Crude birth rate and death rate, Age specific birth rate and death rate. Single and Double sampling plan their OC, AQL, | CO1 CO2 CO2 CO2 | | | | | | | |
| | C Unit 2 A B C Unit 3 A | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and u-chart. Comparison between control charts for variables and control charts for attributes. Analysis of patterns on control chart, estimation of process capability. Techniques of Analysis Crude birth rate and death rate, Age specific birth rate and death rate. Single and Double sampling plan their OC, AQL, LTPD, AOQ, AOQL, ASN, ATI functions with | CO1 CO2 CO2 CO2 CO3 | | | | | | | |
| | C Unit 2 A B C Unit 3 A B | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and u-chart. Comparison between control charts for variables and control charts for attributes. Analysis of patterns on control chart, estimation of process capability. Techniques of Analysis Crude birth rate and death rate, Age specific birth rate and death rate. Single and Double sampling plan their OC, AQL, LTPD, AOQ, AOQL, ASN, ATI functions with graphical interpretation. | CO1 CO2 CO2 CO2 CO3 CO3 | | | | | | | |
| | C Unit 2 A B C Unit 3 A | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and u-chart. Comparison between control charts for variables and control charts for attributes. Analysis of patterns on control chart, estimation of process capability. Techniques of Analysis Crude birth rate and death rate, Age specific birth rate and death rate. Single and Double sampling plan their OC, AQL, LTPD, AOQ, AOQL, ASN, ATI functions with graphical interpretation. Use and interpretation of Dodge and Romig's sampling | CO1 CO2 CO2 CO2 CO3 | | | | | | | |
| | C Unit 2 A B C Unit 3 A B | variation. Statistical Control Charts. Construction and Statistical basis of 3-σ Control charts, Rational Sub-grouping. Control Charts Control charts for variables: X-bar & R-chart, X-bar & s-chart. Control charts for attributes: np-chart, p-chart, c-chart and u-chart. Comparison between control charts for variables and control charts for attributes. Analysis of patterns on control chart, estimation of process capability. Techniques of Analysis Crude birth rate and death rate, Age specific birth rate and death rate. Single and Double sampling plan their OC, AQL, LTPD, AOQ, AOQL, ASN, ATI functions with graphical interpretation. | CO1 CO2 CO2 CO2 CO3 CO3 | | | | | | | |



| В | Problems thereof for weighted and unweighted index numbers including Laspeyre's, Paasche's, Edgeworth- Marshall and Fisher's. | CO4 |
|--------------|---|-----|
| С | Chain index numbers, conversion of fixed based to chain- based index numbers and vice-versa. | CO4 |
| Unit 5 | Consumer price index numbers | |
| А | Consumer price index numbers. | CO5 |
| В | Compilation of indices, base shifting, splicing and deflating of index numbers. | CO5 |
| С | Index of industrial and agriculture production, usage and limitations of index numbers. | CO6 |
| Mode of | Theory | |
| examination | | |
| Weightage | CA 250/. ESE 750/ | |
| Distribution | CA:25%; ESE:75% | |
| Text book/s* | Montgomery, Douglas, C, Introduction to Statistical Quality Control, John Wiley & Sons. M. Jeya Chandra, Statistical Quality Control, CRC Press. | |
| Other | 1.Eugene Lodewick Grant, Richard S. Leavenworth, | |
| References | Statistical Quality Control, McGraw-Hill. | |

| РО | PO | PO | PO | PO | PO | PO | PO | PO | PO | РО | PO | PSO | PSO | PSO |
|----------|----|-----|-----|-----|----|-----|----|-----|----|----|-----|-----|-----|-----|
| CO | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 1 | 2 | 3 |
| MDA116.1 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 3 | |
| MDA116.2 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 3 | |
| MDA116.3 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 3 | |
| MDA116.4 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 3 | |
| MDA116.5 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 3 | |
| MDA116.6 | | 2 | 1 | 2 | | 1 | | 3 | | | 3 | 3 | 3 | |
| Average | | 2.0 | 1.0 | 2.0 | | 1.0 | | 3.0 | | | 3.0 | 3.0 | 3.0 | |