

Vidya Pratishthan's Kamalnayan Bajaj Institute of Engineering and Technology, Baramati

Department of Artificial Intelligence and Data Science

S.Y. B. Tech Syllabus 2025-26 (As per NEP 2020)

Syllabus: Multidisciplinary Minor w. e. f. AY: 2025- 2026

Multidisciplinary Minor in Artificial Intelligence and Data Science

| Course | Courses Name | Teaching Scheme | | | Examination Scheme and Marks | | | | | | | | Credits | | |
|---------|------------------------------------|--------------------|----|-----|------------------------------|-----|-----|----|----|----|-------|----|---------|-----|-------|
| Code | | ТН | PR | TUT | Activit y | ISE | ESE | TW | PR | OR | Total | ТН | PR | TUT | Total |
| AI24052 | Data Processing and Analysis | 2 | 2 | - | 20 | 20 | 50 | 20 | | | 110 | 2 | 1 | | 3 |

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Vidya Pratishthan's

Kamalnayan Bajaj Institute of Engineering and Technology, Baramati (Autonomous Institute)

AI24052- Data Processing and Analysis

Teaching Scheme: Theory: 2 Hours/Week Practical: 2 Hour/Week

Credits 03

Examination Scheme: Activity:20 Marks ISE: 20 Marks ESE: 50 Marks

Term Work: 20 Marks

Prerequisites: Python Programming

Course Objectives:

- To understand the need of Data Science
- To understand computational statistics in Data Science
- To provide a comprehensive knowledge of data science using Python.
- To learn the essential concepts of data analytics and data visualization.

Course Outcomes (COs): The students will be able to learn:

CO1: Apply basic data manipulation techniques using Pandas.

CO2: Identify and apply the need and importance of pre-processing techniques.

CO3: Implement data visualization using visualization tools in Python programming.

CO4: Apply various algorithms and evaluate their performance using metrics.

Course Contents

Unit I Introduction to Data Science and Pandas Basics (06 Hours)

Data science: Definition and importance of data science in various industries. Overview of the data science process and the role of a data scientist. Datafication, and data science lifecycle. Ethical considerations and challenges in data science.

Getting Started with Pandas: Overview of Pandas library and its architecture. Introduction to data structures: Series and DataFrames. Indexing, selection, and filtering data. Basic operations: sorting, ranking, reindexing, and handling missing data.

Unit II Statistical Inference & Data Wrangling (6 Hours)

Statistical Inference: Measures of central tendency: Mean, median, mode, and their application in data analysis. Measures of dispersion: Variance, standard deviation, and range. Introduction to Bayes' Theorem and its relevance to data science. Pearson Correlation and its role in understanding relationships between variables.

Data Wrangling: Combining and merging datasets using Pandas. Techniques for reshaping data (pivoting, melting). Handling overlapping data, removing duplicates, and replacing values. Transforming data for analysis (scaling, encoding, and feature extraction).

Unit III Plotting, Visualization & Exploratory Data Analysis (EDA) (6 Hours)

Plotting & Visualization: Importance of data visualization in data science. Overview of different types of data visualizations: Line plots, bar charts, histograms, scatter plots. Introduction to **Matplotlib** and **Seaborn** for data visualization in Python. Plot customization: Titles, labels, colors, legends, and annotations.

Exploratory Data Analysis (EDA): Visualizing distributions, relationships, and trends in the data. Using heatmaps, pairplots, and correlation matrices for EDA. Identifying patterns, outliers, and potential data issues through visualization.

Unit IV Machine Learning Basics & Model Evaluation (6 Hours)

Introduction to Machine Learning: Overview of machine learning: Types of learning (supervised

vs unsupervised). Introduction to classification and regression tasks. Clustering algorithms: K-Means and hierarchical clustering. Evaluation metrics: Accuracy, precision, recall, F1-score.

Model Evaluation & Selection: Cross-validation and its importance in model evaluation. Overfitting and underfitting: Understanding and mitigating these issues. Introduction to Ridge regression for regularization.

Text Books:

- 1. David Dietrich, Barry Hiller, "Data Science and Big Data Analytics", EMC Education services, Wiley publication, 2012, ISBN0-07-120413-X.
- 2. Wes McKinney, "Python for Data Analysis", O'REILLY, ISBN:978-1-449-31979-3, 1st edition, October 2012.
- 3. Rachel Schutt & O'neil, "Doing Data Science", O'REILLY, ISBN:978-1-449-35865-5, 1st edition, October 2013.

Reference Books:

- 1. Joel Grus, "Data Science from Scratch: First Principles with Python", O'Reilly Media, 2015
- 2. Matt Harrison, "Learning the Pandas Library: Python Tools for Data Munging, Analysis, and Visualization, O'Reilly, 2016.
- 3. Chirag Shah, "A Hands-On Introduction to Data Science", Cambridge University Press, (2020), ISBN: 978-1-108-47244-9.
- 4. Wes McKinney, "Python for Data Analysis", O'Reilly media, ISBN: 978-1-449-31979-3.
- 5. Trent Haunk, "Scikit-learn Cookbook", Packt Publishing, ISBN: 9781787286382

E-Resources:

- 1. https://onlinecourses.nptel.ac.in/noc21 cs69/preview
- 2. https://nptel.ac.in/courses/106106179
- 3. https://onlinecourses.swayam2.ac.in/imb23 mg64/preview

List of Assignments

- 1.Perform the following operations using Python on any open source dataset (e.g., data.csv)
 - Import all the required Python Libraries.
 - Locate open source data from the web (e.g., https://www.kaggle.com). Provide a clear description of the data and its source (i.e., URL of the web site).
 - Load the Dataset into pandas dataframe.
 - Data Preprocessing: check for missing values in the data using pandas isnull(), describe() function to get some initial statistics. Provide variable descriptions. Types of variables etc. Check the dimensions of the data frame.
- 2. Load the open source dataset () and apply statistical inference.
 - Calculate and interpret measures of central tendency (mean, median, mode) and dispersion (variance, standard deviation, range) to summarize data.
 - Explore Pearson correlation to assess relationships between variables.
 - Apply Bayes' Theorem to solve probability-based problems in data analysis. Load the open source dataset ()
- 3. Load the open source dataset (e.g.csv) and perform data wrangling with pandas.
 - Combine and merge multiple datasets using various Pandas functions.

- Reshape data using pivoting and melting techniques to restructure data for analysis.
- Clean data by handling duplicates, overlapping data, and performing value replacements.
- Scale numerical data and encode categorical variables to prepare datasets for statistical analysis or machine learning.
- 4. Load the any open source dataset(e.g. csv) and with Matplotlib and seaborn libraries in python.
 - Create a various plots such as line plots, bar charts, histograms, and scatter plots.
 - Customize plots by adding titles, labels, colors, legends, and annotations to enhance readability and understanding.
 - Use visualizations to identify patterns, outliers, and potential data issues in datasets, facilitating deeper insights into the data for decision-making.
 - Find the correlation between variables using suitable plot.
- 5. Implement a simple unsupervised machine learning model (e.g., K-Means clustering and evaluate its performance using cross-validation.
- 6. Implement a supervised machine learning model (e.g., Logistic Regression or Linear Regression).
 - Train the model and evaluate its performance using accuracy or R-squared (for regression).
- 7. Implement k-fold cross-validation on a regression or classification model (e.g., Linear Regression or Logistic Regression).
 - Compare the results of cross-validation with a single train-test split.
- 8. Choose a classification dataset (e.g., Titanic dataset or Breast Cancer dataset).
 - Implement model evaluation metrics for classification (k-means or logistic regression).
 - Evaluate a classification model using accuracy, precision, recall, and F1-score.