

## NIRMA UNIVERSITY

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| <b>Institute:</b>            | Institute of Technology     |
| <b>Name of Programme:</b>    | BTech (CSE)                 |
| <b>Course Code:</b>          | 3CS110ME24                  |
| <b>Course Title:</b>         | Federated Learning          |
| <b>Course Type:</b>          | Disciplinary Minor-Elective |
| <b>Year of Introduction:</b> | 2024-25                     |

| L | T | Practical Component |    |   | C |
|---|---|---------------------|----|---|---|
|   |   | LPW                 | PW | W |   |
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### Course Learning Outcomes (CLO):

At the end of the course, the students will be able to –

1. explain the fundamentals of federated learning (BL2)
2. make use of techniques of federated learning for developing various applications (BL3)
3. list real-world applications and use cases of federated learning
4. discuss the privacy and security considerations in federated learning. (BL4) (BL6)

| Unit     | Contents   | Teaching Hours<br>(Total 45) |
|----------|--|------------------------------|
| Unit-I   | <b>Introduction:</b> Empirical Probability, Theoretical Probability, Joint Probability, Bayes' Theorem, Descriptive Statistics, Measure of Center, Measure of Variability, Measure of Position, Data visualization, supervised and unsupervised learning   | 05                           |
| Unit-II  | <b>Regression Techniques:</b> Basic concepts and applications of Regression, Simple Linear Regression – Gradient Descent and Normal Equation Method, Multiple Linear Regression, Non-Linear Regression, Linear Regression with Regularization, Hyper-parameters tuning, Loss Functions, Evaluation Measures for Regression Techniques, Artificial neural network, Perceptron Learning, Activation Functions, Multilayer Perceptrons. | 12                           |
| Unit-III | <b>Introduction to Federated Learning:</b> Concept of federated learning, Motivations and advantages, Federated learning as a solution, Current development in federated learning.   | 08                           |
| Unit-IV  | <b>Distributed Machine learning:</b> Introduction to DML, scalability, privacy in DML, privacy-preserving gradient descent. Horizontal federated learning, architecture of HFL, Vertical federated learning, architecture of VFL   | 06                           |
| Unit-V   | <b>Federated Learning Algorithms:</b> Federated Averaging, Federated Stochastic Gradient Descent (FSGD), Federated Learning with Differential Privacy, Other federated learning algorithms   | 04                           |
| Unit-VI  | <b>Privacy and Security in Federated Learning:</b> Differential privacy and federated learning, Secure aggregation and encryption techniques, Threat models and mitigations, Real-world privacy breaches and lessons learned   | 06                           |
| Unit-VII | <b>Application:</b> Healthcare and medical research, Finance and fraud detection, Mobile and IoT applications, Federated learning in federated industries  | 04                           |

**Self-Study:**

The self-study contents will be declared at the commencement of the semester. Around 10% of the questions will be asked from self-study contents

**Suggested Readings/ References:**

1. Qiang Yang Yu, Federated Learning, Morgan and Claypool
2. Muhammad Habib, Federated Learning Systems, Springer
3. Heiko Ludwig, Nathalie Baracaldo, Federated Learning: A Comprehensive Overview of
4. Methods and Applications, Springer
5. Roozbeh Razavi-Far, Boyu Wang, et al., Federated and Transfer Learning: (Adaptation, Learning, and Optimization), Springer

**Suggested List of Experiments:**

| Sr. No. | Title   | Hours |
|---------|---|-------|
| 1       | Introduction to Python and Libraries <ul style="list-style-type: none"> <li>o Set up a Python environment with popular libraries (e.g., NumPy, Pandas, Scikit-Learn).</li> <li>o Load a dataset and perform basic data manipulation tasks.</li> <li>o Explore Jupyter Notebooks for interactive coding.</li> </ul>                      | 02    |
| 2       | Data Preprocessing and Visualization <ul style="list-style-type: none"> <li>o Clean and preprocess a real-world dataset.</li> <li>o Visualize the data using Matplotlib and Seaborn.</li> <li>o Handle missing values, outliers, and feature scaling.</li> </ul>  | 02    |
| 3       | Supervised Learning - Regression <ul style="list-style-type: none"> <li>o Build a simple linear regression model using Scikit-Learn.</li> <li>o Train and evaluate the model on a regression dataset.</li> <li>o Plot the regression line and assess model performance.</li> </ul>  | 02    |
| 4       | Setting Up a Federated Learning Environment <ul style="list-style-type: none"> <li>o Install and configure the necessary libraries and tools for federated learning (e.g., TensorFlow, PyTorch).</li> <li>o Set up a basic federated learning environment.</li> <li>o Train a simple federated model on a synthetic dataset.</li> </ul> | 02    |
| 5       | Federated Averaging Algorithm <ul style="list-style-type: none"> <li>o Implement the Federated Averaging (FedAvg) algorithm.</li> <li>o Use FedAvg to train a basic model across decentralized data sources.</li> <li>o Evaluate the model's performance and compare it to a centralized model.</li> </ul>                              | 04    |
| 6       | Implementing Federated Stochastic Gradient Descent (FSGD) <ul style="list-style-type: none"> <li>o Implement Federated Stochastic Gradient Descent (FSGD).</li> <li>o Train a model using FSGD and compare its convergence with standard SGD.</li> <li>o Discuss the benefits and drawbacks of FSGD.</li> </ul>                         | 04    |
| 7       | Differential Privacy in Federated Learning <ul style="list-style-type: none"> <li>o Implement Federated Learning with Differential Privacy (DP).</li> <li>o Explore the impact of varying privacy parameters on model accuracy.</li> <li>o Discuss the trade-offs between privacy and utility.</li> </ul>                               | 02    |
| 8       | Secure Aggregation and Encryption Techniques  | 04    |

- Implement secure aggregation techniques (e.g., secure sum) in federated learning.
  - Encrypt and decrypt model updates for privacy.
  - Compare the performance of secure aggregation with non-secure methods.
- 9      Federated Learning on Heterogeneous Data Sources                          04
- Simulate federated learning on datasets with varying distributions.
  - Explore techniques for handling non-IID (non-Independently and Identically Distributed) data.
  - Discuss strategies to mitigate issues with heterogeneous data sources.
- 10     Federated Transfer Learning    04
- Implement federated transfer learning to leverage pre-trained models.
  - Fine-tune a pre-trained model on decentralized data sources.
  - Evaluate the transfer learning model's performance.