

| Subject Code | Subject Name      | Credits |
|--------------|-------------------|---------|
| 26CS609      | EDGE INTELLIGENCE | 4       |

### Course Objectives:

1. To design and deploy AI models on resource-constrained edge devices.
2. To explore optimization techniques for efficient edge AI systems.
3. To analyse real-world applications of edge intelligence.
4. To study emerging trends like federated learning and TinyML.

### Learning Objectives:

After completion of the course the students will be able to:

1. Analyse edge computing architectures and paradigms.
2. Design ML models for resource-constrained edge devices.
3. Apply optimization techniques for efficient edge deployment.
4. Evaluate security, privacy, and distributed learning methods.
5. Develop real-time edge AI applications.

### Unit I (12 Hrs.)

**Foundations of edge intelligence:** Cloud, Fog and Edge computing evolution - Edge AI: Concept and Definition -Edge-cloud trade-offs (latency, bandwidth, privacy) - Edge architecture - Resource constraints (compute, memory, energy) - Real-time edge applications

### Unit II (12 Hrs.)

**Edge-oriented machine learning:** ML pipeline for edge systems - Lightweight models (MobileNet, SqueezeNet) -TinyML and embedded inference - On-device vs cloud training - Multi-objective optimization - Frameworks (TensorFlow Lite, ONNX Runtime)

### Unit III (12 Hrs.)

**Optimization and deployment:** Quantization and pruning techniques - Knowledge distillation methods - Edge-specific Neural Architecture Search - Graph optimization and operator fusion - Edge deployment pipeline - Runtime inference optimization strategies

### UNIT IV (12 Hrs.)

**Distributed edge intelligence & security:** Federated and split learning - Communication-efficient training - Differential privacy and encryption - Adversarial and poisoning attacks - Blockchain-based trust management - Fault tolerance and energy-aware AI

### UNIT V (12 Hrs.)

**Edge ai systems design:** Edge AI system design - Task partitioning and scheduling - Latency and energy modeling - Resource management and scaling - Performance benchmarking - Real-world deployment cases

### Reference Books:

1. Pete Warden & Daniel Situnayake, TinyML: Machine Learning with TensorFlow Lite on Arduino and Ultra-Low-Power Microcontrollers, 2019.
2. Qiang Yang et al., Federated Learning: Collaborative Machine Learning with Privacy, 2019.
3. Yu Shi et al., Edge AI: On-Demand Accelerated Deep Learning Inference, 2020.
4. Rajkumar Buyya et al., Cloud, Edge, and Fog Computing: Technologies and Applications, 2021.
5. Ian Goodfellow, Yoshua Bengio, Aaron Courville, Deep Learning, 2016.