# **ASSIGNMENT-3**

**Objective:** Build on the image classification model from Assignment 2 by implementing **active learning strategies** to enhance model performance and efficiency in training.

# **Key Requirements:**

### 1. CNN Model Implementation (Baseline):

- Retain the CNN architecture designed in Assignment 2 with at least two convolutional layers, ReLU activation, and max pooling.
- Continue experimenting with architectural components (e.g., number of layers, filter sizes) and hyperparameters (e.g., learning rate, batch size) to optimize performance.
- Use a suitable dataset for training (e.g., CIFAR-10, MNIST, Fashion-MNIST).

# 2. Optional Pretrained Model Extension:

- Utilize a pretrained model (e.g., ResNet, VGG, etc.) from the PyTorch model zoo.
- Fine-tune the pretrained model for your chosen dataset and compare its performance with the custom CNN.

#### 3. Active Learning Integration:

Implement active learning strategies to enhance data efficiency by selecting the most informative samples for training. The strategies should include:

#### Uncertainty Metrics:

- **Least Confidence**: Select samples where the model has the least confidence in its predictions.
- Prediction Entropy: Use entropy to measure uncertainty in predictions.
- Margin Sampling: Choose samples where the difference between the top two predicted probabilities is smallest.

### o Diversity Metrics:

- Cosine Similarity: Ensure diverse selection of samples by minimizing cosine similarity.
- L2 Norm: Evaluate diversity based on the Euclidean distance in feature space.

• **KL Divergence**: Compare the distribution of predicted probabilities to enhance diversity.

## 4. Training and Evaluation:

- o Divide the dataset into labeled and unlabeled subsets for active learning.
- Use the active learning strategies to iteratively select batches of samples for labeling and training.
- Compare the model's performance (e.g., accuracy) using active learning against the baseline model.

### 5. Report and Analyze Results:

- o Report the classification accuracy achieved for:
  - Custom CNN.
  - Pretrained model (if implemented).
  - Active learning-enhanced model.
- Provide insights on how active learning impacts training efficiency and accuracy.
- Highlight the most effective active learning strategy for your dataset and model.