DS5007 Deep Learning Lab 6 - GNN, GAN and Diffusion

Max Marks: 10

Deadline: 11/04/2025, 12:00 PM

Instructions

- 1. Provide well-commented, indented code with meaningful variable names.
- 2. Write the task description in separate text blocks before the corresponding code block.
- 3. Carefully follow the task requirements and use only the specified libraries or approaches.
- 4. Ensure all plots have appropriate axis labels, titles, and legends.
- 5. Submit a single Jupyter Notebook (.ipynb) file named YourName_YourRollNo_Assignment6.ipynb.

Objective:

This lab assignment is designed to give you hands-on experience with three important deep learning models:

- 1. Graph Neural Networks (GNNs)
- 2. Generative Adversarial Networks (GANs)
- 3. Diffusion Models

You will implement, train, and evaluate these models on different datasets.

Tasks & Marks Distribution

Task 1: Graph Neural Network (GNN) Implementation (3 Marks)

- 1. **Dataset:** Use the **Cora dataset**.
- 2. Task: Implement a Graph Convolutional Network (GCN) for node classification.
- 3. Requirements:
 - Load the dataset and preprocess it (normalize adjacency matrix, convert to sparse tensors).
 - o Define a 2-layer GCN model
 - Train the model and report test accuracy.
 - Visualize node embeddings using **T-SNE**.

Task 2: Generative Adversarial Network (GAN) Implementation (3 Marks)

- 1. **Dataset:** Use the **CIFAR-10** dataset.
- 2. Task: Implement a Deep Convolutional GAN (DCGAN) to generate images.
- 3. Requirements:
 - Load and preprocess the dataset (normalize to [-1, 1]).
 - Implement a DCGAN with:
 - Generator (Conv2DTranspose layers).
 - Discriminator (Conv2D layers).
 - o Train the GAN and generate **sample images** after every few epochs.
 - Plot the loss curves for generator and discriminator.

Task 3: Diffusion Model Implementation (4 Marks)

- 1. **Dataset:** Use the **CIFAR-10** dataset.
- 2. Task: Implement a Denoising Diffusion Probabilistic Model (DDPM) for image generation.
- 3. Requirements:
 - Load and preprocess the dataset (rescale to [-1, 1]).
 - o Implement:
 - Forward diffusion process (gradually add Gaussian noise).
 - Reverse diffusion process (U-Net denoising model).
 - Train the model and generate samples at different noise levels.
 - Compare generated images with real images.

Note:

- Include plots, loss curves, and generated samples for each task.
- For GNNs, **Spektral** is recommended (install via pip install spektral).