Build and train a GAN for generating hand-written digits

1. Hyperparameters:

- latent_dim: Size of the random noise vector used as input for the generator.
- hidden_dim: Size of the hidden layers for both the generator and discriminator.
- o image_dim: Flattened size of the MNIST image (28x28 pixels = 784).
- num_epochs: Number of training epochs.
- batch_size: Batch size for training.
- 1r: Learning rate for the Adam optimizer.
- beta1: The exponential decay rate for the first moment estimates in the Adam optimizer.
- 2. **Device Setup**: The code will automatically use a GPU if available.
- 3. **MNIST Dataset**: The MNIST dataset is loaded and normalized to a range of [-1, 1] to match the output range of the generator.

4. Generator Model:

- Takes a random noise vector (latent_dim) as input.
- Uses fully connected layers with ReLU activations and batch normalization.
- The output is a 28x28 image with pixel values between [-1, 1] (tanh activation).

5. **Discriminator Model**:

- o Takes an image (either real or generated) and classifies it as real or fake.
- Uses fully connected layers with LeakyReLU activations.

6. Loss Function:

 Adversarial loss (BCELoss): Binary Cross-Entropy loss is used for both the generator and discriminator to measure how well each can distinguish real from fake images.

7. **Optimizers**: Adam optimizers are used for both the generator and discriminator with a learning rate of 0.0002 and $\beta1$ of 0.5.

8. Training Loop:

- Discriminator: It is trained using both real images (from MNIST) and fake images (generated by the generator).
- Generator: It is trained to generate images that can fool the discriminator into classifying them as real.

9. Image Generation:

- Every 10 epochs, the generator creates 16 fake images, which are saved as .png files to monitor the progress of training.
- At the end of training, the generator creates a final set of images and displays them.
- 10. **Saving the Model**: The trained generator model is saved for later use.