GANs (Generative Adversarial Networks)

1. Architecture:

- Generator: Creates data (e.g., images) from random noise.
- o **Discriminator**: Distinguishes between real and generated data.

2. Training:

- Adversarial Process: Generator and Discriminator train simultaneously, competing against each other.
- o **Goal**: Generator learns to create realistic data to fool the Discriminator.

3. Applications:

 Image generation, data augmentation, super-resolution, and image-to-image translation.

DCGANs (Deep Convolutional GANs)

1. Improvements:

Uses convolutional layers for better image processing.

2. Generator Enhancements:

- Transposed Convolutions: Upsample noise to high-res images.
- o Batch Normalization: Stabilizes training.
- **ReLU/Tanh Activations**: Used for better performance.

3. Discriminator Enhancements:

- Convolutional Layers: Downsample images.
- Leaky ReLU: Helps with training stability.
- Batch Normalization: Stabilizes training.

4. Benefits:

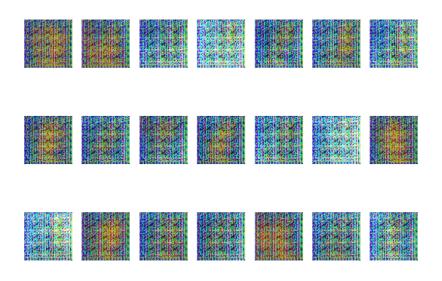
- Produces more realistic images.
- o Improved training stability.

Summary

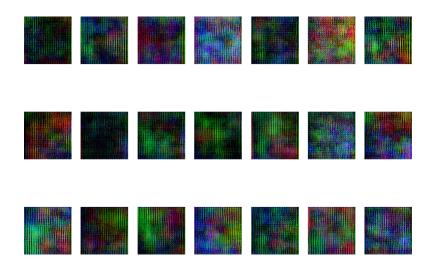
- GANs: Two networks (Generator and Discriminator) with adversarial training.
- DCGANs: Enhanced GANs using convolutional layers for higher quality and stability in image generation.

Results:

1) Epoch 1



2)Epoch 4



3)Epoch 7









































