What are GANs? Generative Adversarial Networks (GANs) consist of two competing neural networks: **Generator**: Creates fake data samples. **Discriminator**: Determines whether samples are real or fake. Training is a minimax game, aiming to reach Nash equilibrium. Key Applications of GANs Image Processing & Computer Vision Super-resolution (e.g., SRGAN, ESRGAN) Image synthesis (e.g., DR-GAN, PG-GAN) Texture synthesis & object detection Sequential Data NLP (e.g., RankGAN, IRGAN) Music generation (e.g., RNN-GAN, ORGAN) Other Fields Medical imaging Data augmentation Style transfer Advantages of GANs ✓ Generates realistic samples without explicit probability models. ✓ Works with high-dimensional & complex data. Can be trained without labeled data. ✓ Used for unsupervised & semi-supervised learning. Challenges & Limitations Mode collapse – Generator produces limited variety. **Training instability** – Hard to balance Generator & Discriminator. Computationally expensive – Requires large datasets & computing power.