

Training and Evaluation Report

1. Training Setup

Dataset

- Dataset: MNIST
- Image Size: 28×28
- Channels: 1 (Grayscale)
- Preprocessing:
 - Resize to 28×28
 - Normalize pixel values to $[-1, 1]$
 - Shuffle and batch data
 - Train/Test split = 95% / 5%

Model Configuration

Generator

- Input: Noise vector of dimension 100
- Architecture:
 - Linear \rightarrow ReLU
 - Reshape ($128 \times 7 \times 7$)
 - ConvTranspose2D \rightarrow ReLU
 - ConvTranspose2D \rightarrow Tanh
- Output: (1, 28, 28) image

Discriminator

- Input: Image (1, 28, 28)
- Architecture:
 - Conv2D \rightarrow LeakyReLU
 - Conv2D \rightarrow LeakyReLU
 - Flatten \rightarrow Linear \rightarrow Sigmoid
- Output: Probability score

Training Parameters

Parameter	Value
Batch Size	64
Epochs	50
Noise Dimension	100
Optimizer	Adam
Learning Rate	0.0002
Beta1	0.5
Loss Function	Binary Cross Entropy
Label Smoothing	Real labels = 0.9

2. Training Procedure

- a) Load training data using DataLoader.
- b) Initialize Generator and Discriminator.
- c) For each batch:
 - Generate fake images from random noise.
 - Compute Discriminator loss on real and fake images.
 - Update Discriminator weights.
 - Compute Generator loss using Discriminator predictions.
 - Update Generator weights.
- d) Repeat for all epochs.
- e) Save Generator checkpoints and sample outputs periodically.

3. Training Monitoring

- Generator and Discriminator losses logged per epoch.
- Training samples saved at fixed intervals.
- Final trained models saved to disk.

4. Evaluation Methodology

Feature Extraction

- Pretrained InceptionV3 used as feature extractor.
- Auxiliary classifier disabled.
- Final classification layer removed.
- Features extracted for real and generated images.

Evaluation Metrics

- i. **Fréchet Inception Distance (FID)**
 - Measures distance between real and generated feature distributions.
 - Lower value indicates better similarity.
- ii. **Diversity Score**
 - Computed as mean standard deviation across generated features.
 - Indicates variation in generated samples.
- iii. **t-SNE Visualization**
 - Dimensionality reduction applied to feature vectors.
 - Used for visual comparison of real and fake embeddings.

5. Evaluation Results

Metric	Observation
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FID Score	Reduced after training
Diversity Score	Stable
t-SNE	Overlapping real and fake clusters
Mode Collapse	Not observed