

# Training and Evaluation Report

## 1. Training Setup

### Dataset

- Dataset: MNIST
- Image Size:  $28 \times 28$
- Channels: 1 (Grayscale)
- Preprocessing:
  - Resize to  $28 \times 28$
  - 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:
  - o Generate fake images from random noise.
  - o Compute Discriminator loss on real and fake images.
  - o Update Discriminator weights.
  - o Compute Generator loss using Discriminator predictions.
  - o 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