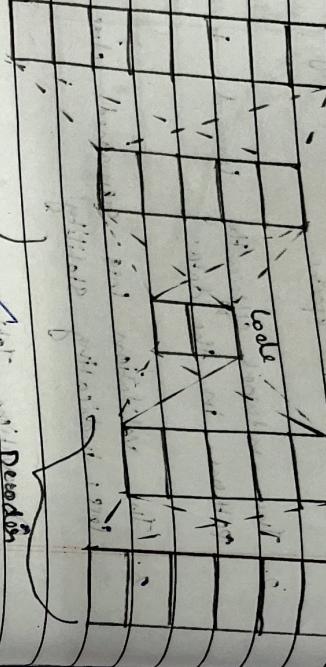


17-10-25

Auto Encoder Architecture

Input Output



Decoder

Aims:

- To perform image compression on the MNIST dataset, using an Autoencoder Neural Network.
- Train the model to reconstruct a noisy image given compressed representations.
 - Build an Autoencoder for dimensionality reduction.
 - Evaluate compression efficiency and reconstruction accuracy.

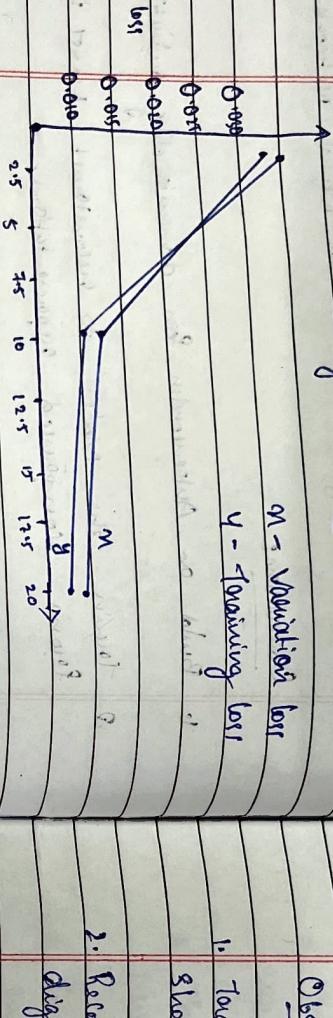
PSEUDO CODE:

- Load MNIST dataset.
- Normalise pixel values.
- Define Encoder(Dense layer \rightarrow Compressed code).
- Define Decoder(Dense layer \rightarrow reconstructed image).
- Combine encoder + decoder = Autoencoder.
- Compile Optimizer - 'adam', loss = 'mse'.
- Train model (x, train, X, train)
- Evaluate \rightarrow reconstruction loss.
- Visualize Original vs reconstructed images.
- Save compressed features.

Observation Table:

Epoch	Training loss	Validation loss
1	0.032	0.031
10	0.010	0.011
20	0.009	0.009

Training & Validation loss Over Epoch



Observation:

1. Training and Validation loss decreased steadily showing efficient learning.
2. Reconstructed image closely matches original digits.

3. The encoder effectively compressed high dimensional images into compact latent features.

Result:

Result:

The Autoencoder successfully compressed MNIST images with minimal info loss, achieving efficient reconstruction.

Epochs	10	20
Training loss	0.032	0.010
Validation loss	0.031	0.011

Result: