

Day 10:
Autoencoders

Introduction to Autoencoders

What

Special type of neural network - Designed for Unsupervised learning

Primarily - dimensionality reduction and data compression, feature reduction

Essential principle

Learning to compress and effectively represent input data without specific labels

Reduces the complexity of data, which reduces irrelevant data and improves performance

Working

Using a two-fold structure that consists of an encoder and a decoder

Encoder:

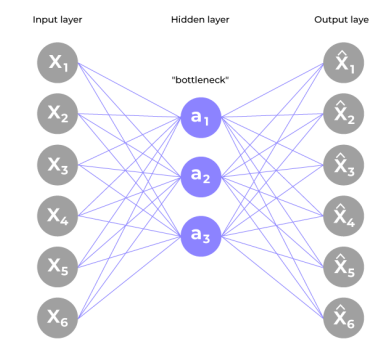
Transforms the input data into a reduced-dimensional representation (a.k.a. latent space)

Decoder:

rebuilds the initial input

This process also defines essential features to gain meaningful patterns

An encoder, decoder, and bottleneck layer



Encoder:

Input layer takes raw input data

Hidden layers progressively reduce the dimensionality of the input

Captures important features and patterns

Decoder:

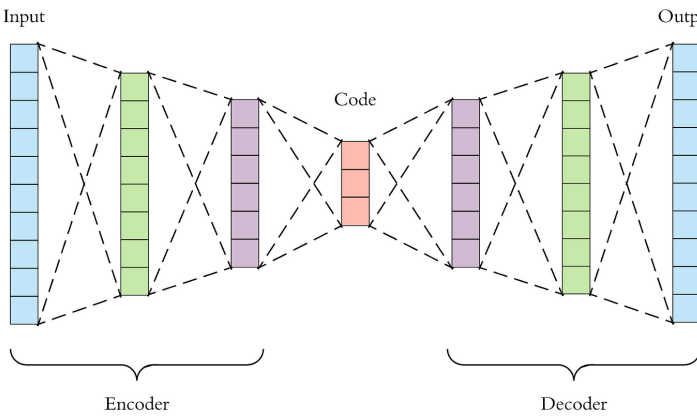
Bottleneck layer: takes the encoded representation and expands it back to the dimensionality of the original input

Hidden layers progressively increase the dimensionality and aim to reconstruct the original input

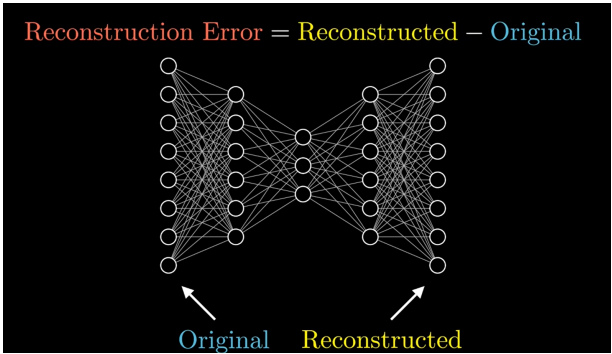
Output layer produces the reconstructed output, which ideally should be as close as possible to the input data

These results backpropagate the neural network in the form of the loss function

Architecture



Loss function used during training is the reconstruction loss



Measures the difference between the input and the reconstructed output

During training the autoencoder learns to minimize the reconstruction loss

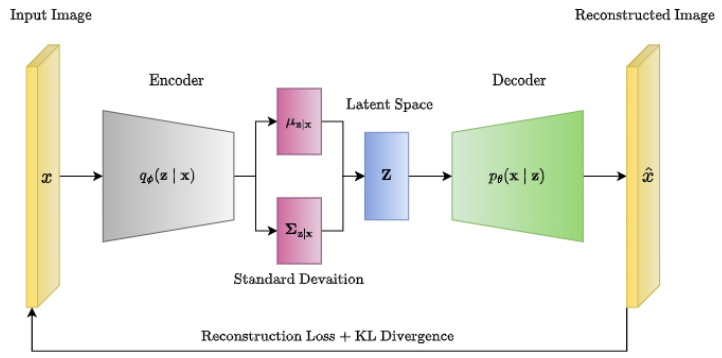
Which forces the network to capture the most important features in the bottleneck layer

Introduction to Variational Autoencoders (VAEs)

It is a more advanced version of the Autoencoder in a way that it provides a statistical manner for describing the samples of the dataset in latent space

In a VAE: the encoder outputs a probability distribution in the bottleneck layer instead of a single output value

Architecture of Variational Autoencoder



The encoder network takes raw input data and transforms it into a probability distribution within the latent space

Decoder takes the sampled point from the latent distribution and reconstructs it back into data space

The loss function in VAE consists of two terms: Reconstruction Loss & Regularization term

Reconstruction Loss: Compels the model to accurately reconstruct the input

Regularization Term: Makes the latent space stick to the chosen distribution

These parameters are iteratively adjusted to make the VAE learn to encode input data into meaningful latent space representation

Hands-On Project: Variational Autoencoders for Image Compression