

# Statistical Data Analysis

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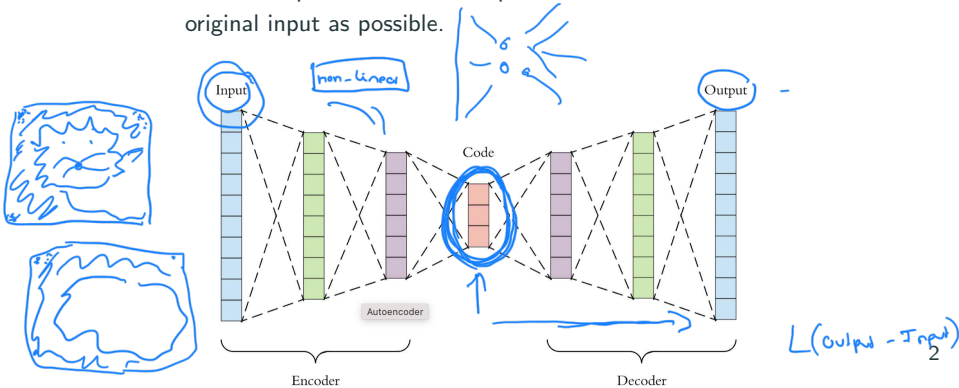
# Autoencoders

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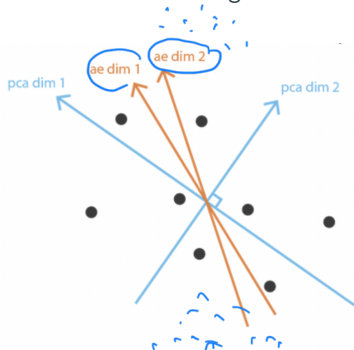
## Snapshot information

- unsupervised artificial neural network (feed forward)
- Two steps:
  - **Encoder:** learns how to efficiently compress and encode data
  - **Decoder:** learns how to reconstruct the data back from the reduced encoded representation to a representation that is as close to the original input as possible.



# PCA vs Autoencoders

- PCA is essentially a linear transformation but Auto-encoders are capable of modelling complex non linear functions.
- PCA features are totally linearly uncorrelated while autoencoded features might have correlations
- PCA is faster and computationally cheaper than autoencoders.
- Autoencoder is prone to overfitting due to high number of parameters.  
(though regularization and careful design can avoid this)



# Linear Autoencoders

Set up: have Data  $Y \in \mathbb{R}^{n \times N}$

$$Y_0 = Y - \bar{Y} \mathbf{1}_N^T$$

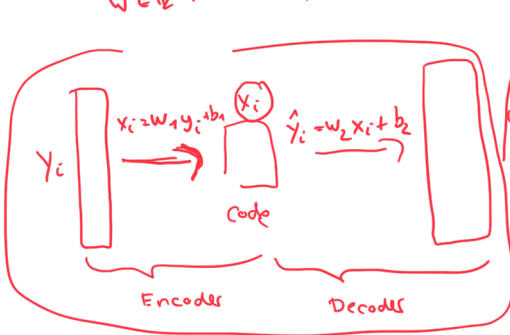
$$\begin{array}{ccc} x_i & = & W^T y_i \\ \uparrow & & \uparrow \\ \mathbb{R}^m & & \in \mathbb{R}^n \\ & & W \in \mathbb{R}^{n \times m} \end{array}$$

$$P_1 = \max_{w_1} w_1^T Y_0 Y_0^T w_1$$

$$\text{s.t. } w_1^T w_1 = 1$$

PCA:

$$Y_0 Y_0^T = \underbrace{P \Lambda P^{-1}}_P = P \Lambda (P^T)$$



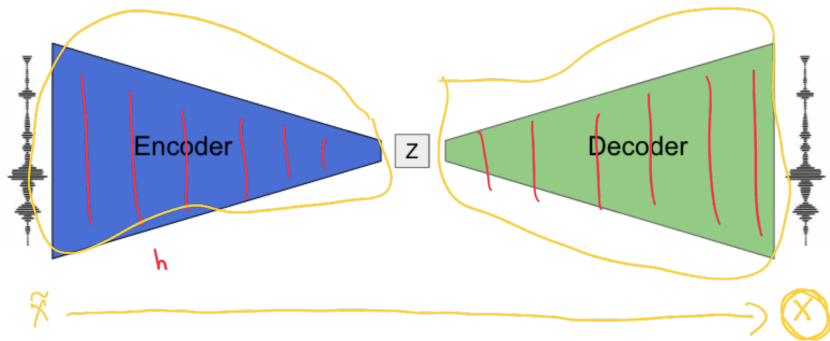
$$\min_{w_1, b_1, w_2, b_2} \| Y - (w_2 (w_1 \odot + b_1 \cdot \mathbf{1}_N^T) + b_2 \cdot \mathbf{1}_N^T) \|$$

$$\min_{w_1, w_2} \| Y_0 - w_2 w_1 Y_0 \|$$





# General Autoencoders



$$\|A\tilde{x} - x\|$$

$$\|f(\tilde{x}) - x\|$$

$$\|g(f(\tilde{x})) - x\|$$



# Regularised Autoencoders

$$\| \quad \| \quad \quad \quad \| \\ L(x, g(f(x))) + \Omega(h, x) \quad (1)$$

where  $h = f(x)$  is the encoder and  $g(h)$  the decoder,  $L$  is a choice of loss function. The regularisation term can have the form

$$\Omega(h, x) = \lambda \sum_i \|\nabla_x h_i\| \quad (2)$$

An autoencoder with this regularisation is known as contractive autoencoder.

# Denoising Autoencoders

Traditionally, autoencoders minimize some function

$$L(x, g(f(x))) \quad (3)$$

while a so called denoising autoencoder (DAE) instead minimizes

$$L(x, g(f(\tilde{x}))) \quad (4)$$

