

Machine Learning and Pattern Recognition Practice

Session V: Autoencoder

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1 Perceptron

2 Dimensionality Reduction with Neural Networks

Perceptron

Parametrized supervised learning

optimization problem in supervised learning

$$\hat{y} = f_w(x)$$

$$\hat{w} = \operatorname{argmin}_w L(y, \hat{y}) = \operatorname{argmin}_w L(y, f_w(x))$$

Dimensionality Reduction with Neural Networks

Dimensionality Reduction



Figure: Dimensionality reduction.

The dimensionality reduction goal is to learn a mapping function $f(x) = z$ that maps input samples x into z .

Autoencoder concept

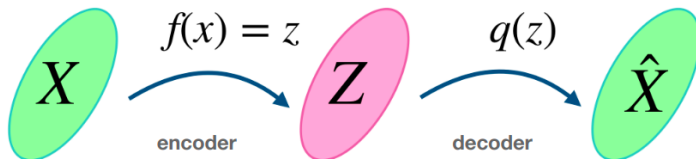
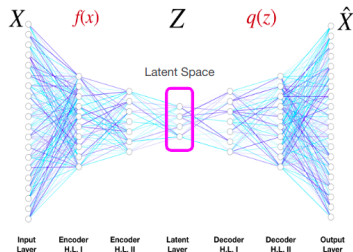


Figure: Encoder and decoder functions.

The autoencoder model is composed by two functions. The encoder $f(x) = z$ maps the input data into a low dimensional latent space. The decoder function reconstructs the samples from the latent space to the original input space.

Autoencoder model

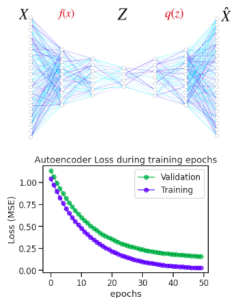


$$z = f(x, \mathbf{W}_f) = \sigma(\mathbf{W}_f \mathbf{x} + \mathbf{b}_f)$$
$$\tilde{x} = q(z, \mathbf{W}_q) = \sigma(\mathbf{W}_q \mathbf{z} + \mathbf{b}_q)$$

Figure: Autoencoder architecture

The autoencoder is a neural network model with one or multiple hidden layers. The hidden layer of the middle is named bottle neck or latent space.

Autoencoder loss function



$$z = f(x, \mathbf{W}_f) = \sigma(\mathbf{W}_f x + \mathbf{b}_f)$$
$$\tilde{x} = q(z, \mathbf{W}_q) = \sigma(\mathbf{W}_q z + \mathbf{b}_q)$$

$$L(x, \hat{x}) = L(x, q(f(x))) = \|x - \hat{x}\|^2$$

Figure: The landscape of activation functions

Dimensionality Reduction

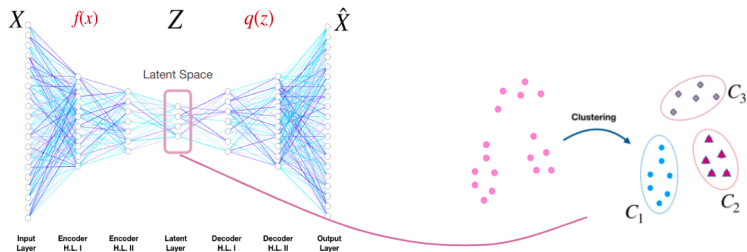


Figure: The landscape of activation functions



Stevens, E., Antiga, L., & Viehmann, T. (2020). Deep learning with PyTorch. Manning Publications.



Shawe-Taylor, J., Cristianini, N. (2004). Kernel methods for pattern analysis. Cambridge university press.