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Relatório de Atividades

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1 Introduction

1 INTRODUCTION

Variational auto-encoder (VAE) model is a stochastic inference and learning algorithm based on variational Bayes (VB) inference proposed by [Kingma and Welling \(2014\)](#). This is a generative technique whose central idea is the development of representations in a low-dimensional latent space that can be mapped back into a realistic-looking image.

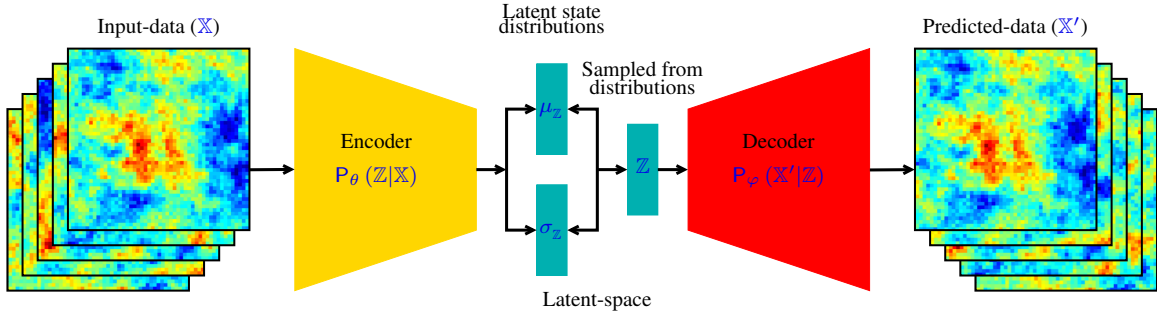
[Higgins et al. \(2016\)](#) introduced the β -VAE, a modification of the original VAE, that introduces an adjustable hyperparameter β to balance latent channel capacity and independence constraints with reconstruction accuracy. They demonstrate that with tuned values of β ($\beta > 1$) the β -VAE outperforms VAE ($\beta = 1$).

2 ZHANG ET AL. (2022)

Variational auto-encoder (VAE) model is a generative network based on variational Bayes (VB) inference proposed by [Kingma and Welling \(2014\)](#).

[Zhang et al. \(2022\)](#) proposed a method to reconstruct porous media based on VAE and Fisher information with good quality and efficiency.

Consider the input data $\mathbb{X} = \{\mathbf{x}^{(i)}\}_{i=1}^N$ consisting of N independent and identically distributed (*i.i.d.*) samples of the continuous (or discrete) variable \mathbf{x} .



$$\mathbb{Z} = \mu_{\mathbb{Z}} + \sigma_{\mathbb{Z}} \cdot \varepsilon, \quad \text{where } \varepsilon \sim \mathbb{N}(0, 1) \quad (1)$$

The Kullback–Leibler divergence (also called relative entropy and I-divergence) is a measure of divergence between two distributions ([Kullback and Leibler, 1951](#); [Csiszar, 1975](#)):

$$\mathcal{D}_{\text{KL}}(f_{\mathbb{P}} || f_{\mathbb{Q}}) = \quad (2)$$

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REFERENCES

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variational auto-encoding, [1](#)