

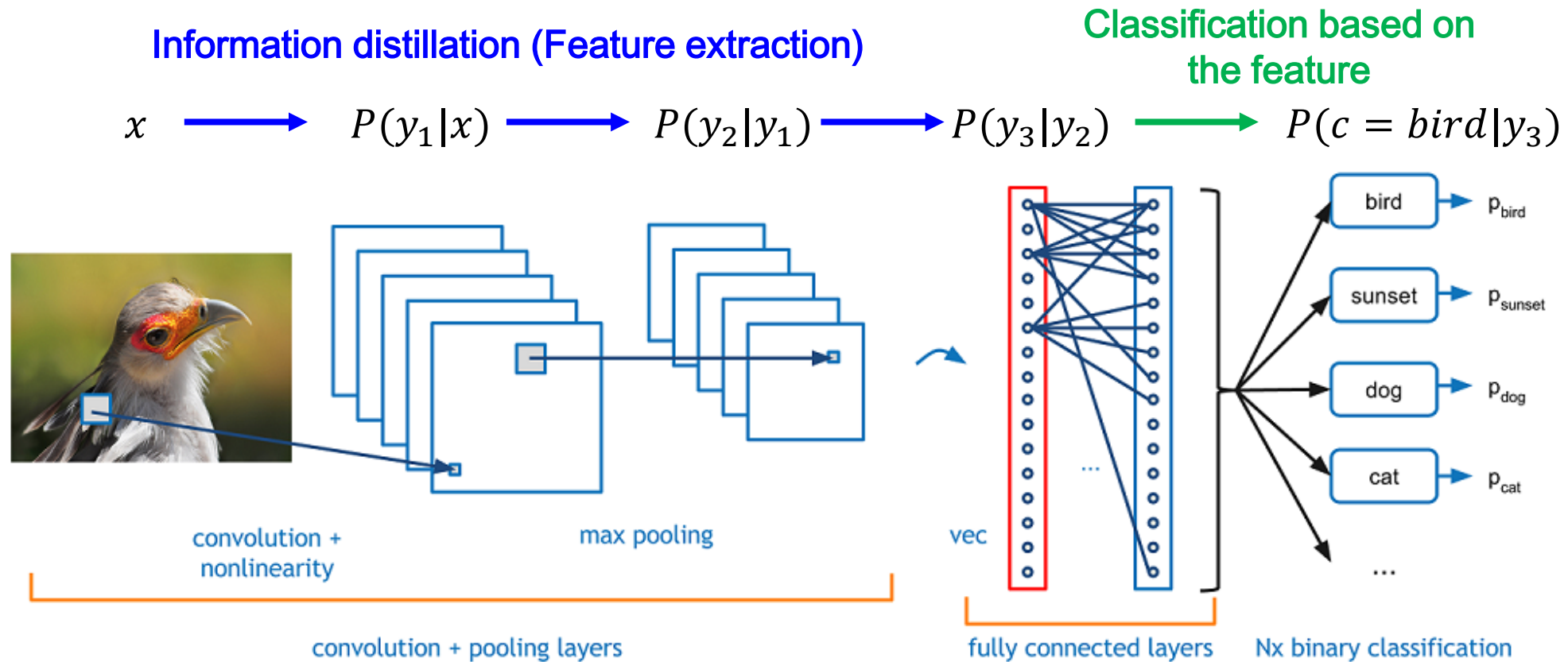
Molecular generative model (2)

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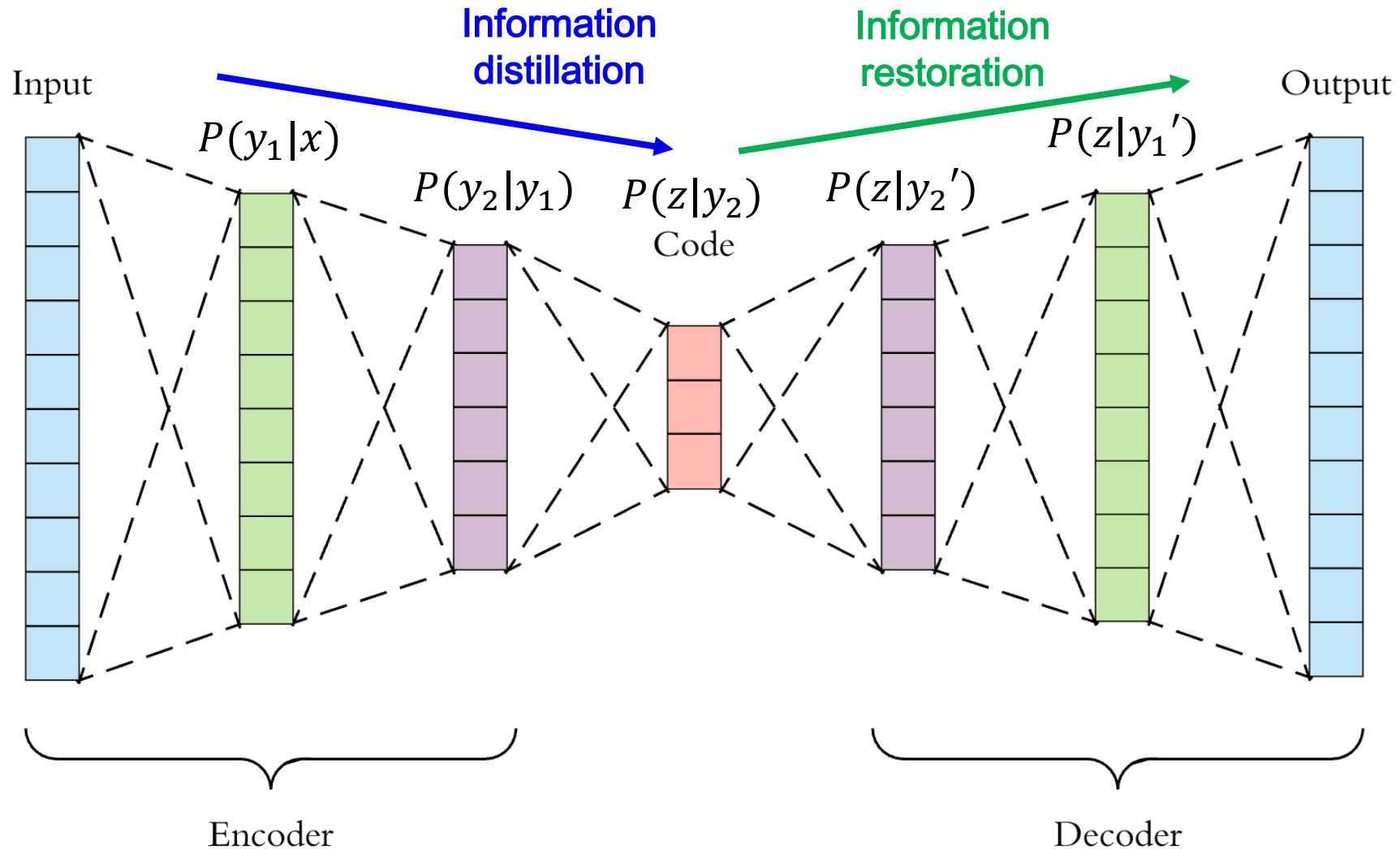
Why does deep learning work so well?

Sufficient statistics



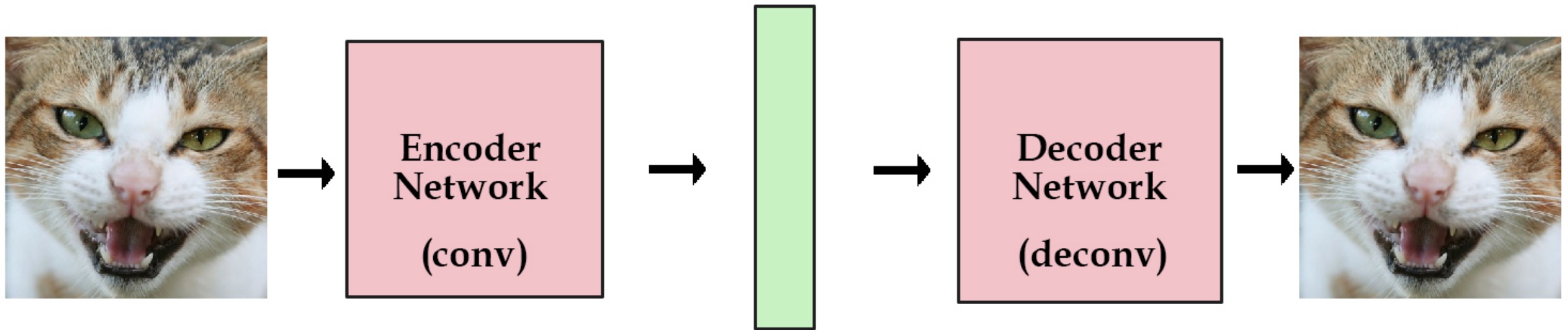
Autoencoder

Can we extract the feature without given labels?



Autoencoder

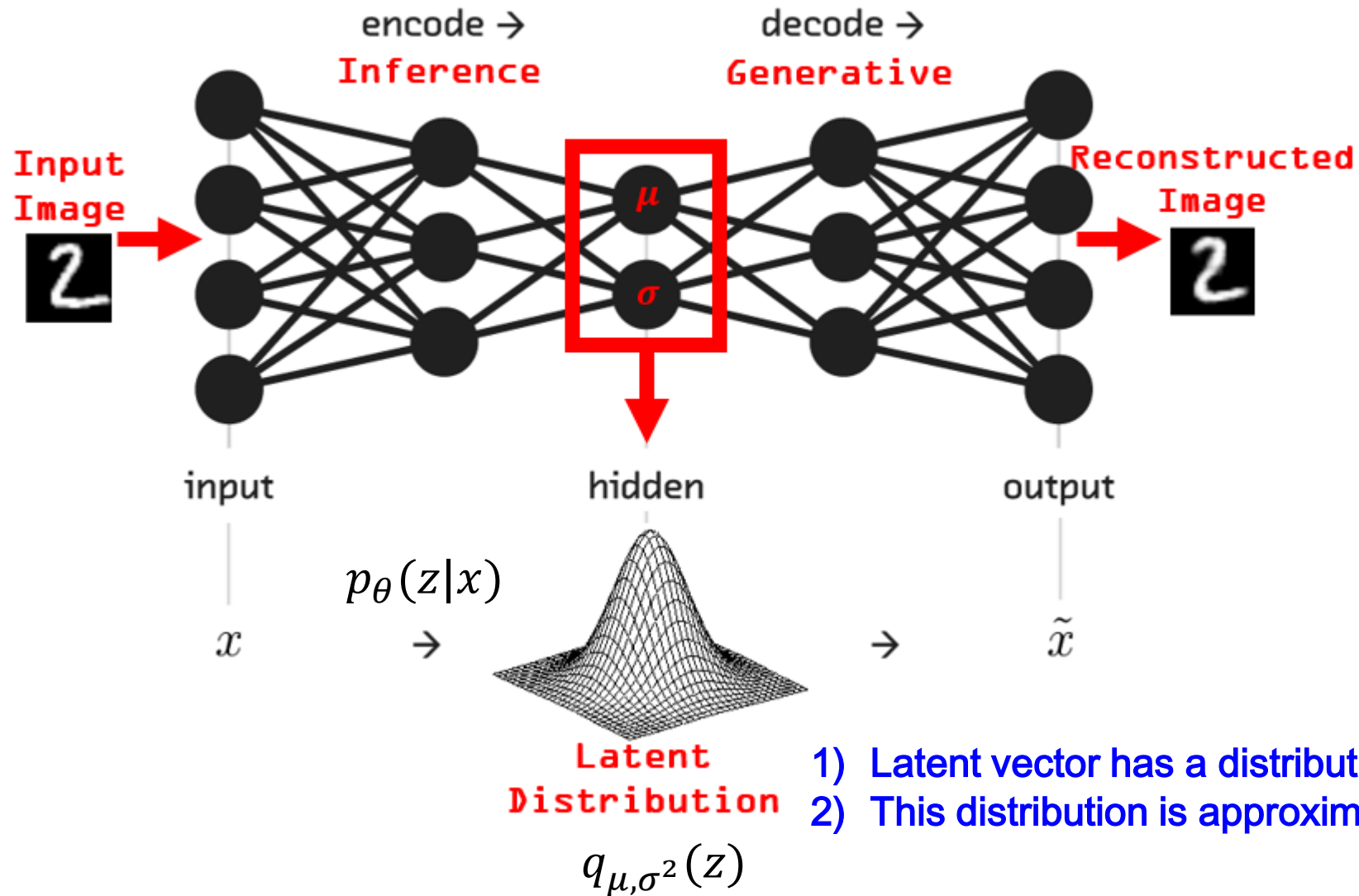
Can we extract the feature without given labels?



latent vector / variables

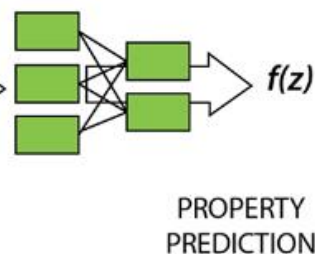
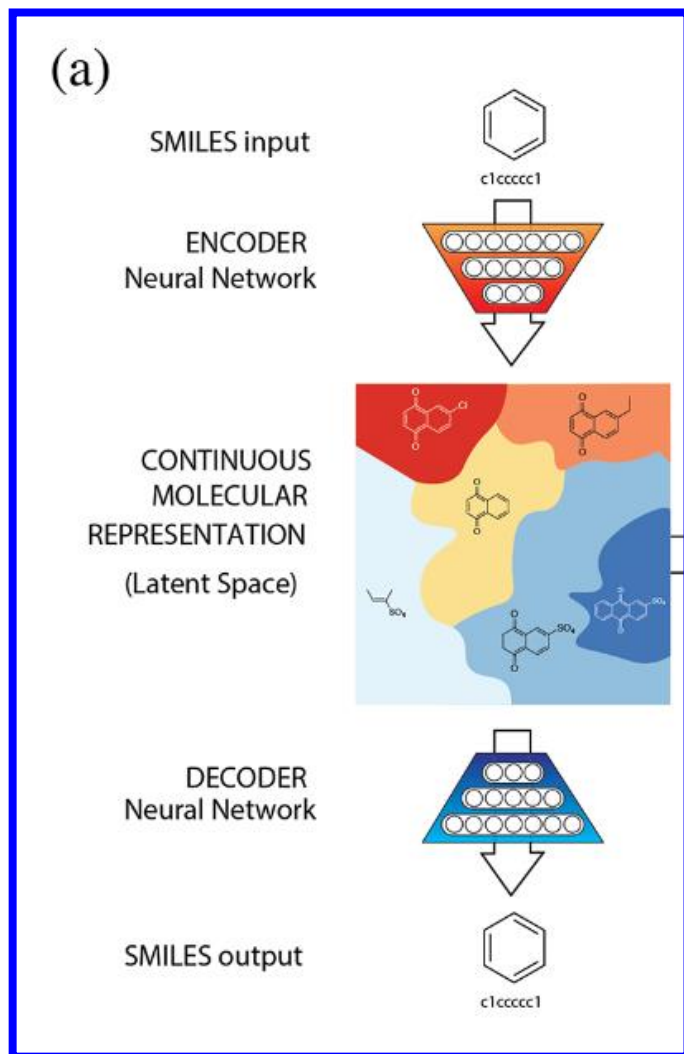
: contains the essential information of given input

Variational autoencoder

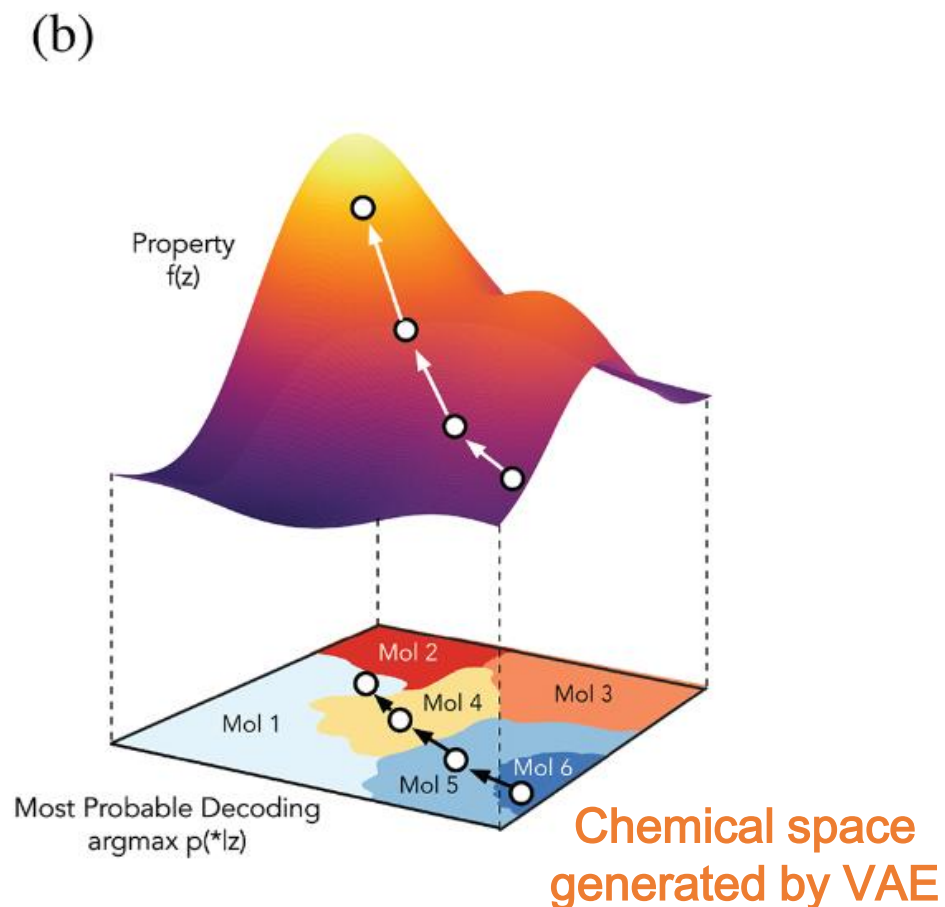


Variational autoencoder

ChemicalVAE



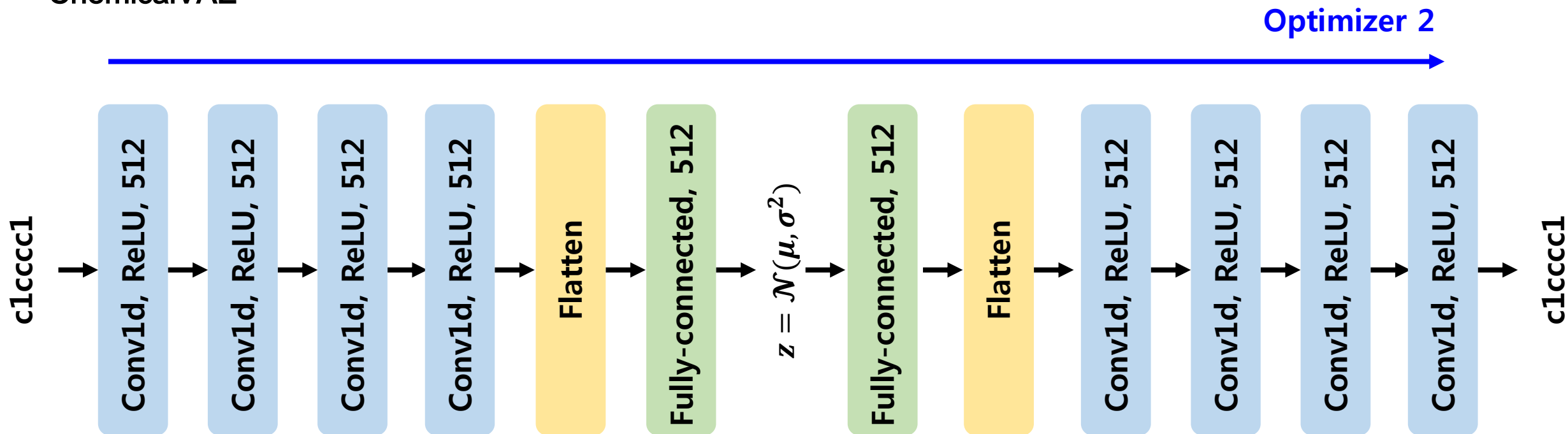
Interest of last practice



Gómez-Bombarelli, Rafael, et al. "Automatic chemical design using a data-driven continuous representation of molecules." *ACS central science* 4.2 (2018): 268-276.

Variational autoencoder

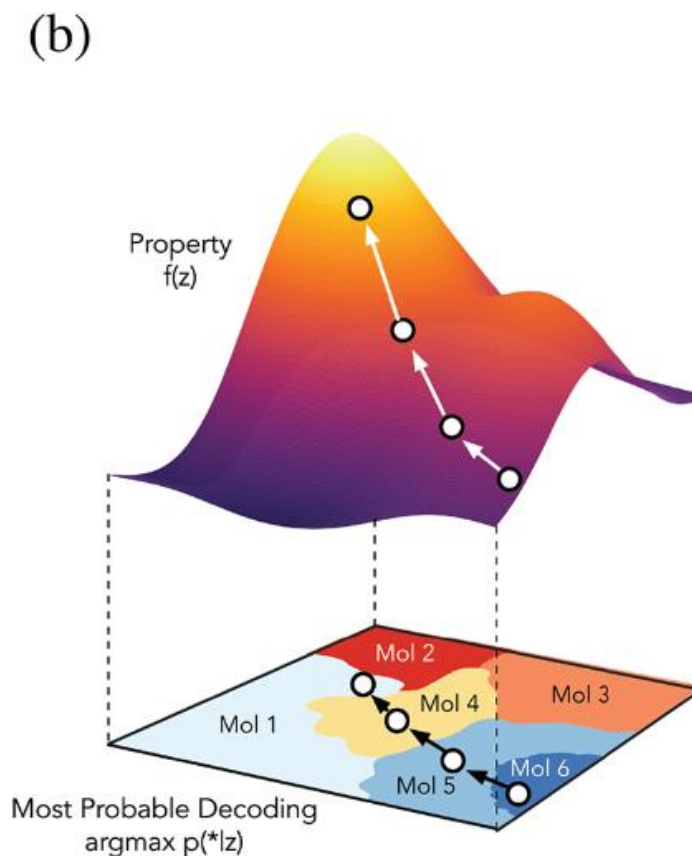
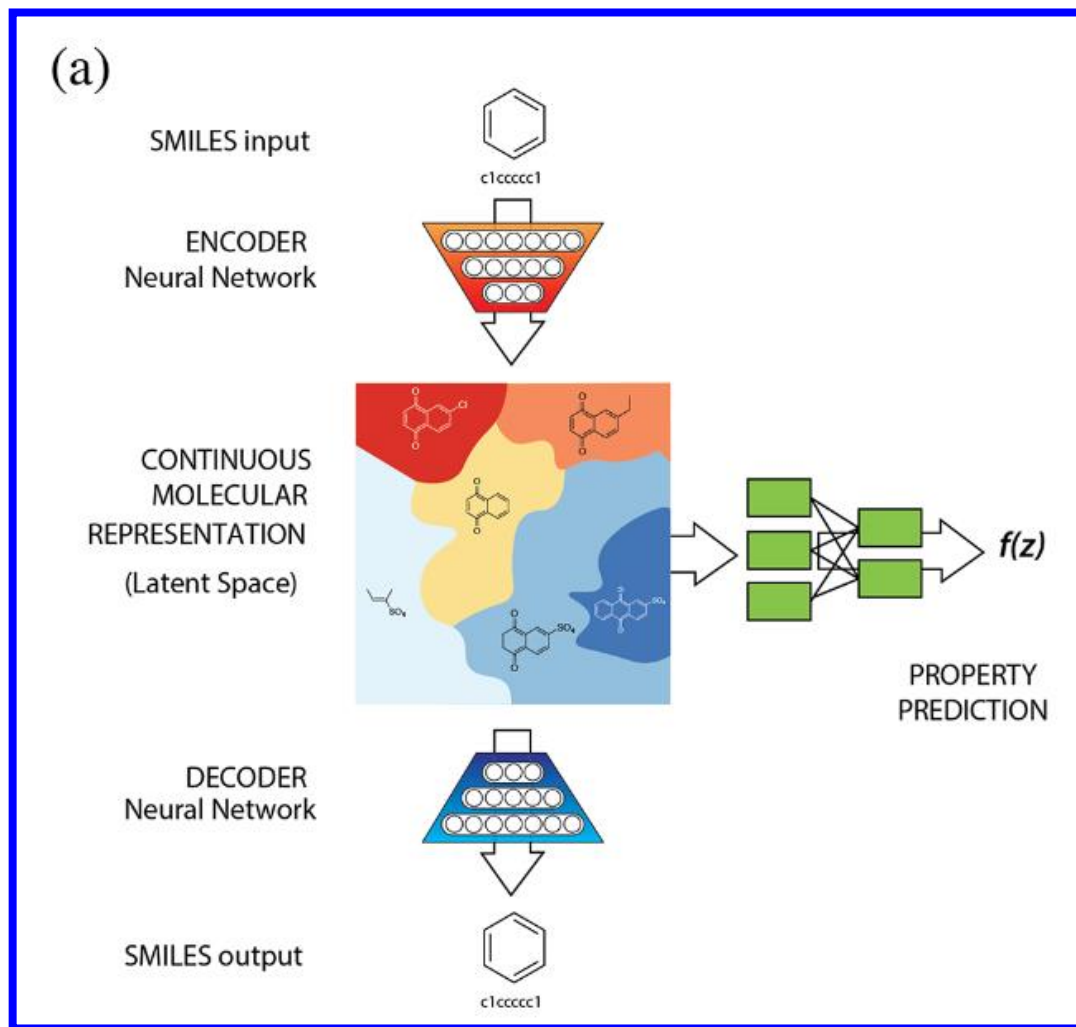
ChemicalVAE



$$\mathcal{L} = \mathcal{L}_{reconstr} = \sum_i \|x_i - \hat{x}_i\|^2$$

Variational autoencoder

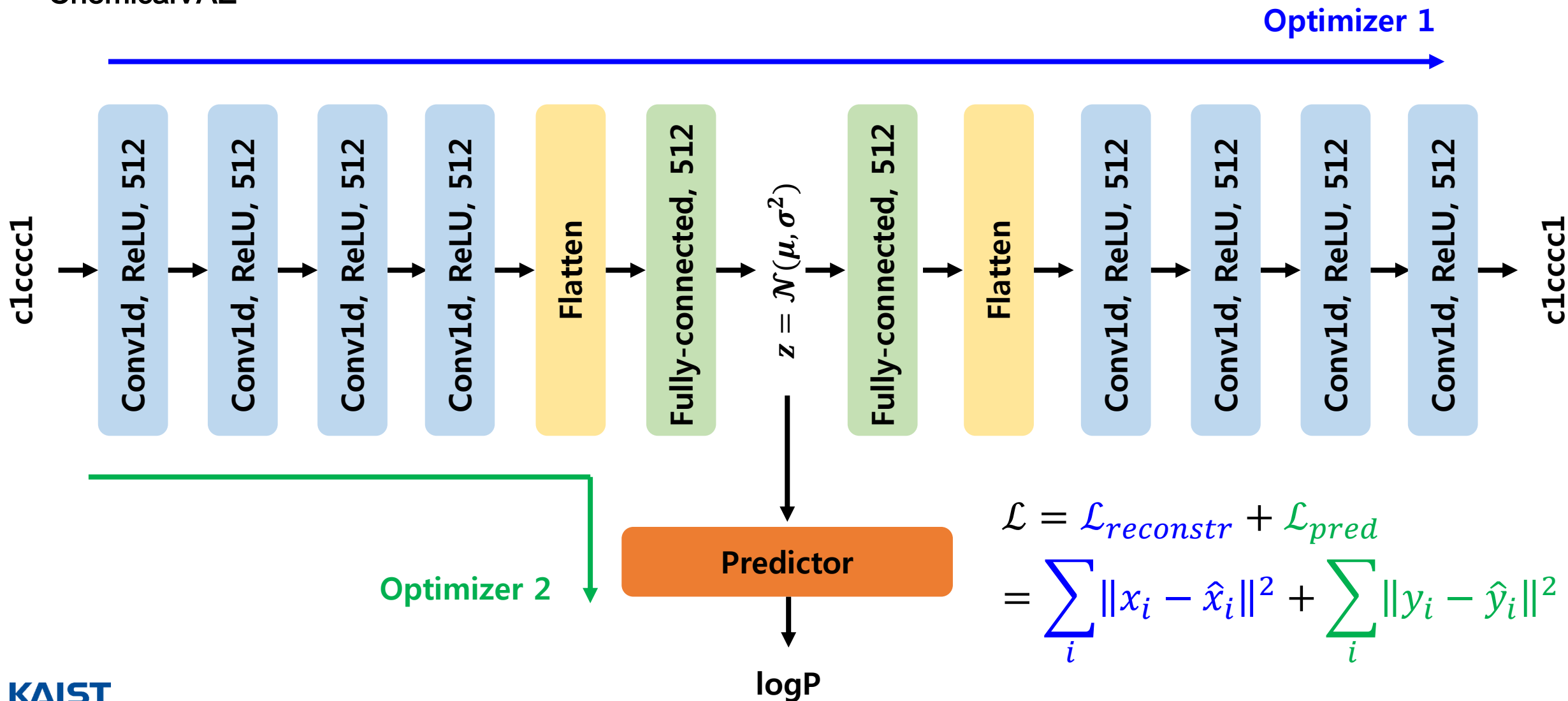
ChemicalVAE



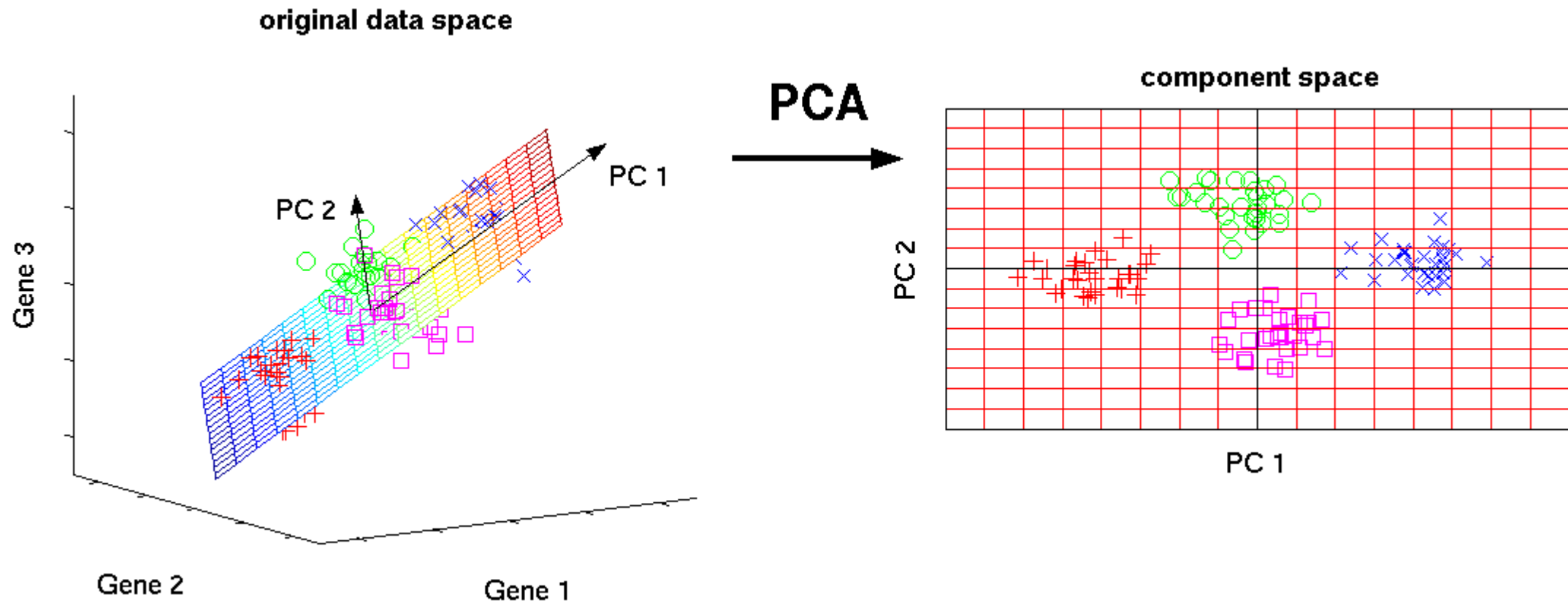
Interest of this practice

Variational autoencoder

ChemicalVAE

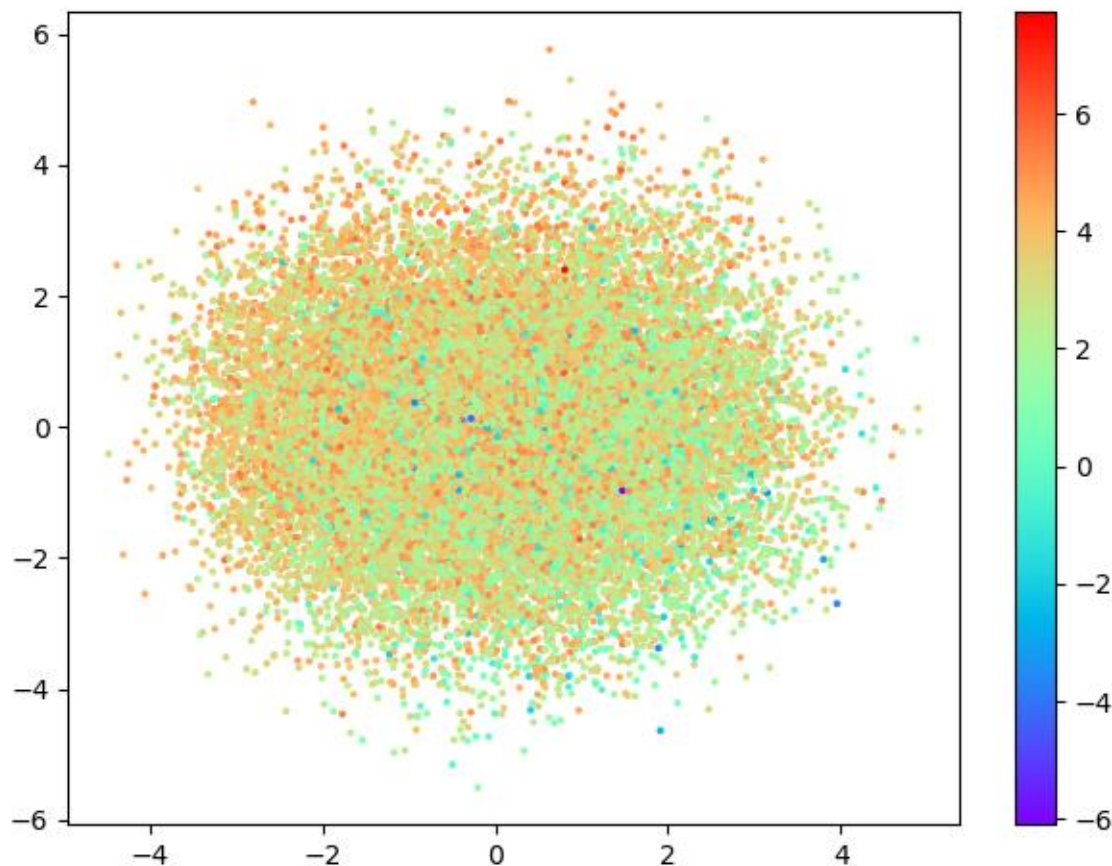


Principal Component Analysis

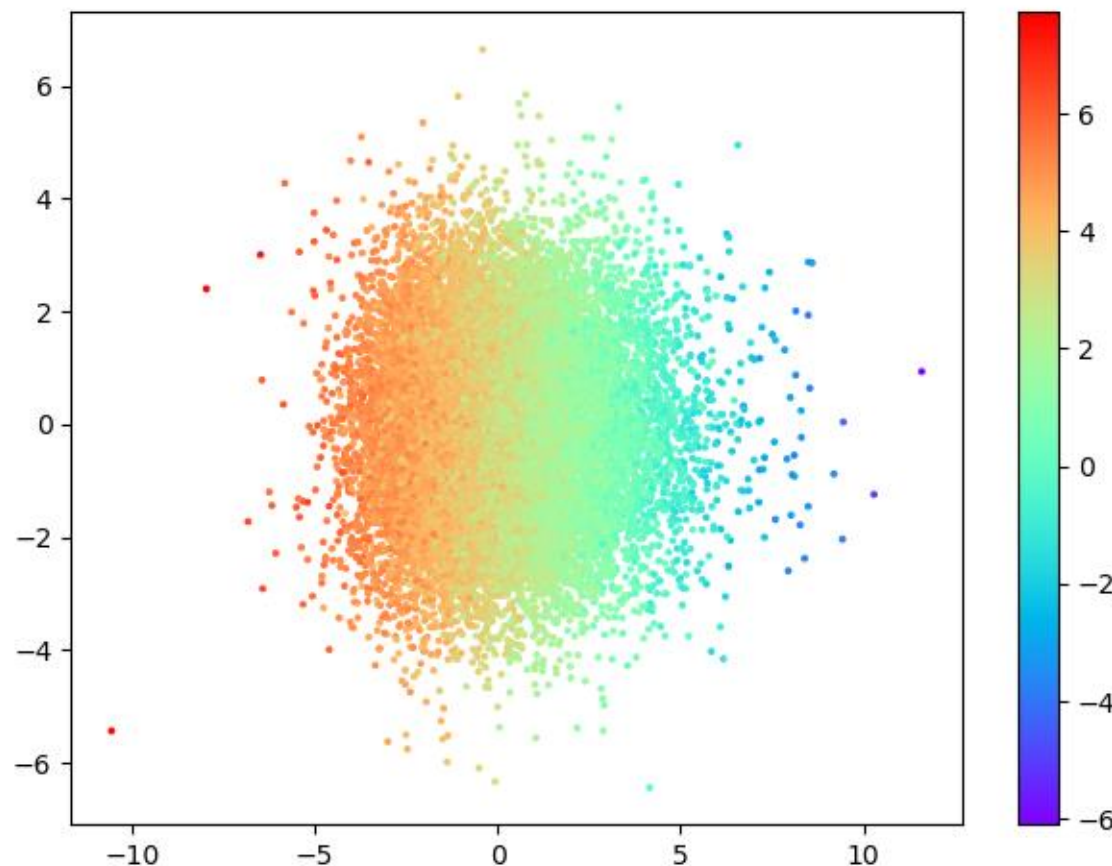


Principal Component Analysis

VAE

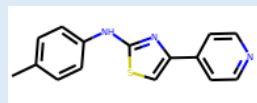
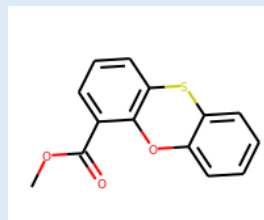
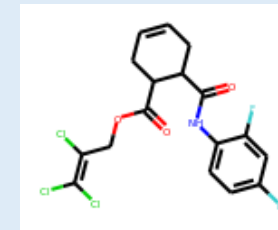
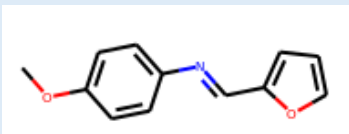


VAE + jointly training



Jointly training makes latent vectors of molecules having similar property closer in the latent space.

Latent space searching



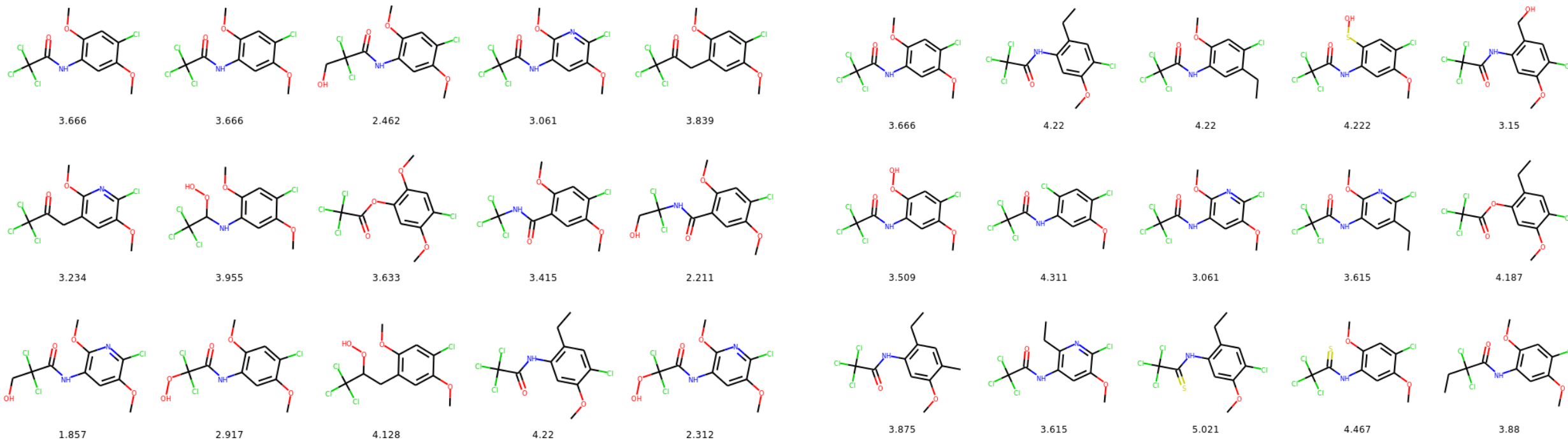
Latent space

$\sigma = 0.1$

Latent space searching

VAE

VAE + jointly training



$\sigma = 0.3$

Latent space searching

VAE

VAE + jointly training

