



pumas<sup>AI</sup>

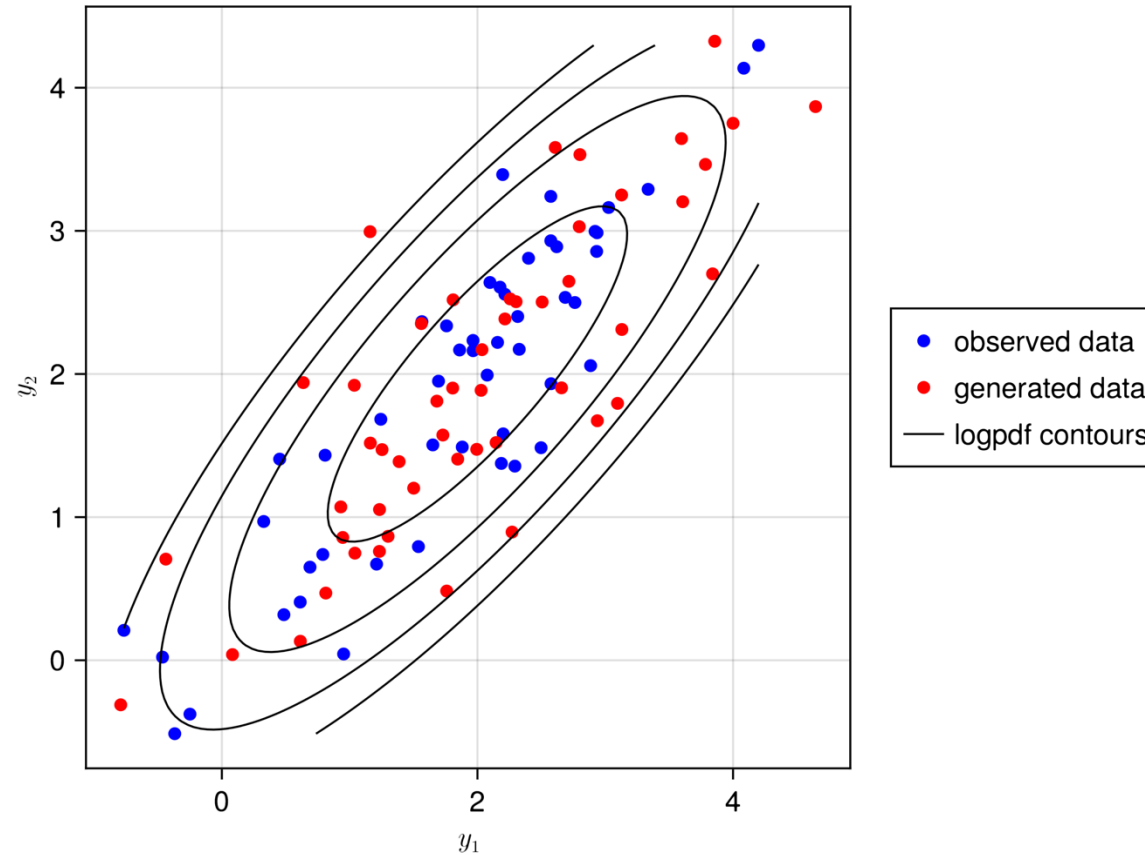
# DeepPumas Embedding models

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# Generative AI

Goal:  
Generated data instinguishable in  
distribution from real data





[www.thispersondoesnotexist.com](http://www.thispersondoesnotexist.com)



# How is that done?

Data is a mix of

- Observed quantities (pixel intensities)
- Unobserved quantities (face, smiling, ...)

We humans learn to extract the unobserved  
GenAI needs to do that too.







# How is that done?

Data is a mix of

- Observed quantities ( $\mathbf{y}$ )
- Unobserved quantities ( $\mathbf{z}$  – “latent variables”)





# Generative models

- Definitions
  - $\mathbf{z}$ : latent variables of dimension  $d$
  - $\mathbf{y}$ : observed response/data
  - $\mathbf{y}_g$ : generated/simulated/synthetic response/data
- Model

$$\begin{aligned}\mathbf{y}_g &= \mathbf{f}(\mathbf{z}) + \boldsymbol{\epsilon} \\ \mathbf{z} &\sim \text{Normal}(0, \mathbf{I}_{d \times d}) \\ \epsilon_i &\sim \text{Normal}(0, \sigma^2)\end{aligned}$$

- Objective: choose  $\mathbf{f}$  such that the distribution of  $\mathbf{y}_g$  is close to the distribution of the observed data  $\mathbf{y}$



# Conditional generative models

- Definitions
  - $\mathbf{z}$ : latent variables of dimension  $d$
  - $\mathbf{x}$ : observed covariates
  - $\mathbf{y}$ : observed response
  - $\mathbf{y}_g$ : generated/simulated/synthetic response
- Model

$$\begin{aligned}\mathbf{y}_g &= \mathbf{f}(\mathbf{z}, \mathbf{x}) + \boldsymbol{\epsilon} \\ \mathbf{z} &\sim \text{Normal}(0, \mathbf{I}_{d \times d}) \\ \epsilon_i &\sim \text{Normal}(0, \sigma^2)\end{aligned}$$

- Objective: choose  $\mathbf{f}$  such that the conditional distribution of  $\mathbf{y}_g \mid \mathbf{x}$  is close to the conditional distribution of the observed data  $\mathbf{y} \mid \mathbf{x}$

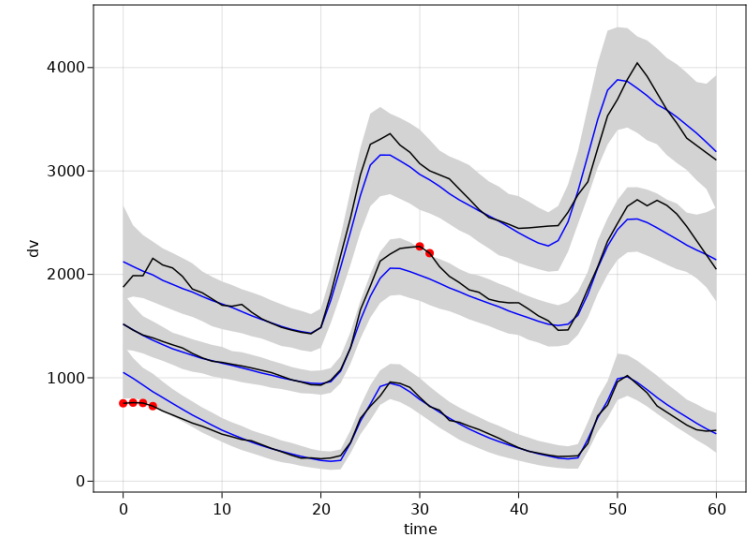


# NLME is Generative AI!

- Definitions
  - $\boldsymbol{\eta}$ : latent variables of dimension  $d$  and covariance matrix  $\boldsymbol{\Omega}$
  - $\boldsymbol{x}$ : observed covariates
  - $\mathbf{dv}$ : observed response
  - $\mathbf{dv}_g$ : generated/simulated/synthetic response
- Model

$$\begin{aligned}\mathbf{dv}_g &= \mathbf{f}_{\boldsymbol{\theta}}(\boldsymbol{\eta}, \boldsymbol{x}) + \boldsymbol{\epsilon} \\ \boldsymbol{\eta} &\sim \text{Normal}(0, \boldsymbol{\Omega}) \\ \epsilon_i &\sim \text{Normal}(0, \sigma^2)\end{aligned}$$

- Objective: choose  $\mathbf{f}_{\boldsymbol{\theta}}$  such that the conditional distribution of  $\mathbf{dv}_g \mid \boldsymbol{x}$  is close to the conditional distribution of the observed data  $\mathbf{dv} \mid \boldsymbol{x}$







# Latent variables = Random effects

## An Introduction to Variational Autoencoders

Diederik P. Kingma, Max Welling

$$p_{\theta}(\mathbf{x}, \mathbf{z}) = p_{\theta}(\mathbf{z})p_{\theta}(\mathbf{x}|\mathbf{z}) \quad (1.14)$$

where  $p_{\theta}(\mathbf{z})$  and/or  $p_{\theta}(\mathbf{x}|\mathbf{z})$  are specified. The distribution  $p(\mathbf{z})$  is often called the *prior distribution* over  $\mathbf{z}$ , since it is not conditioned on any observations.

Maximize the marginal likelihood.

Marginalize over  $\mathbf{z}$

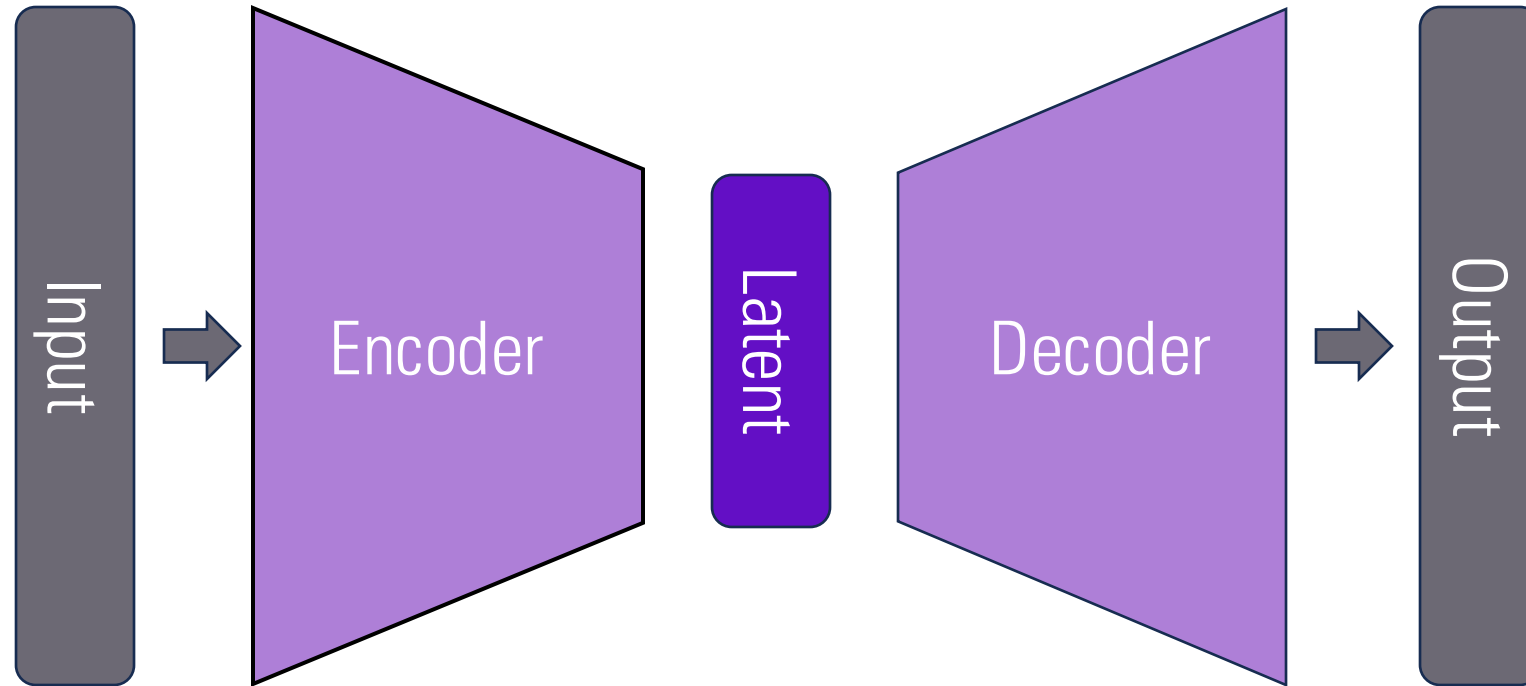
The marginal distribution over the observed variables  $p_{\theta}(\mathbf{x})$ , is given by:

$$p_{\theta}(\mathbf{x}) = \int p_{\theta}(\mathbf{x}, \mathbf{z}) d\mathbf{z} \quad (1.13)$$

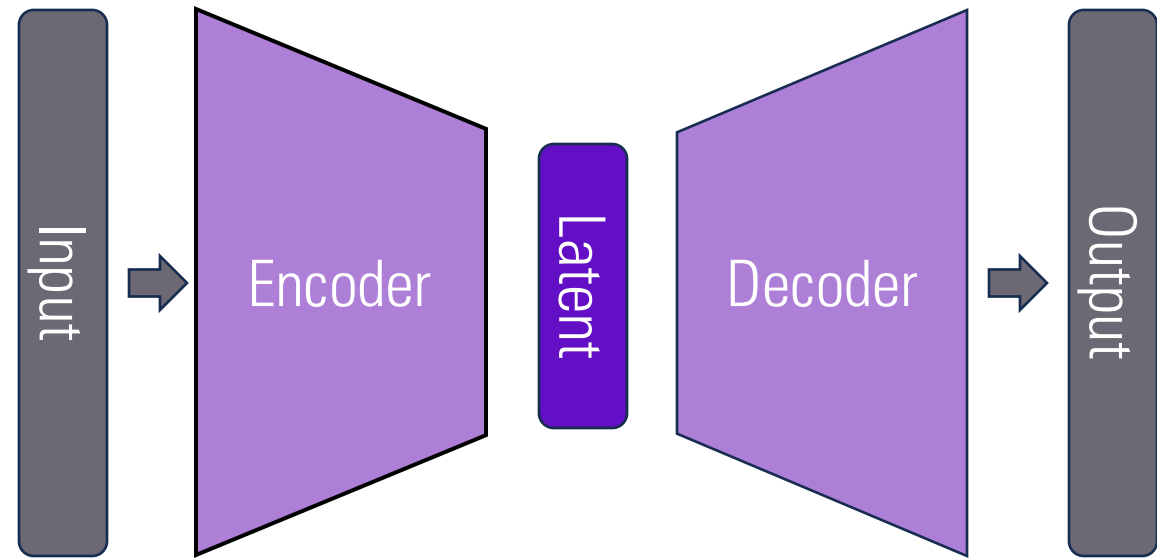
This is also called the (single datapoint) *marginal likelihood* or the *model evidence*, when taken as a function of  $\theta$ .



# Generative AI – Typical anatomy



# NLME as GenAI



Input: Time series of outcomes/concentrations

Output: Time series of outcomes/concentrations

Decoder: The structural NLME (parameter transforms, dynamics, etc.)

Encoder: ???





# What do “latent variables” represent?

NLME:

- Interpretation from structural constraints:  $CL = tvCL \cdot \exp(\eta_{cl})$ .
  - These interpretations can fool us!
- Structural constraints not necessary:  $NN \left( \frac{Central}{Vc}, \eta \right)$



# What do “latent variables” represent?

Image GenAI:

For an image:

- Not pixel-by-pixel intensity

Rather:

- What objects are in the image
- What are the characteristics of the objects
- What are they doing
- What's the style



# What do “latent variables” represent?

Text GenAI:

- Not words

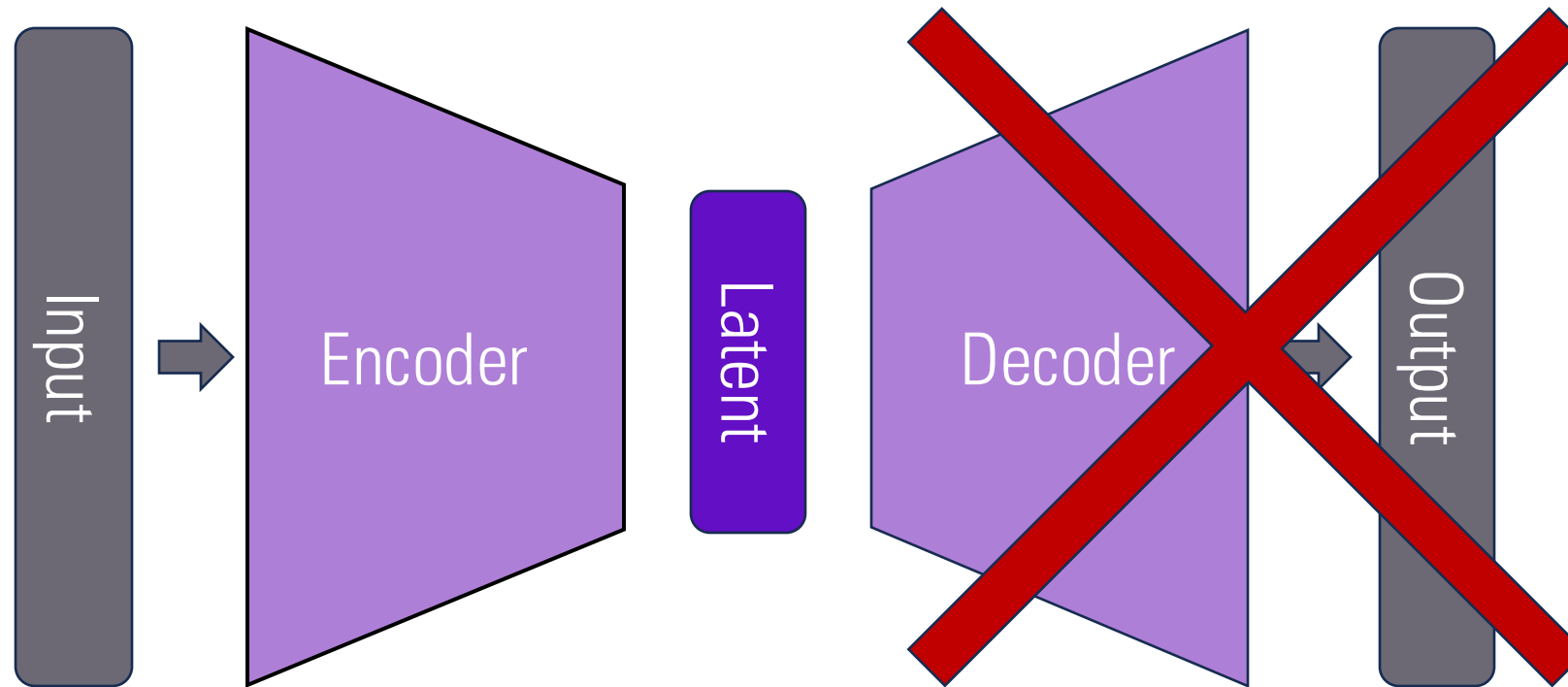
rather:

- Sentiment
- Conveyed information
- Writing style
- Language
- ...





# Embedding models

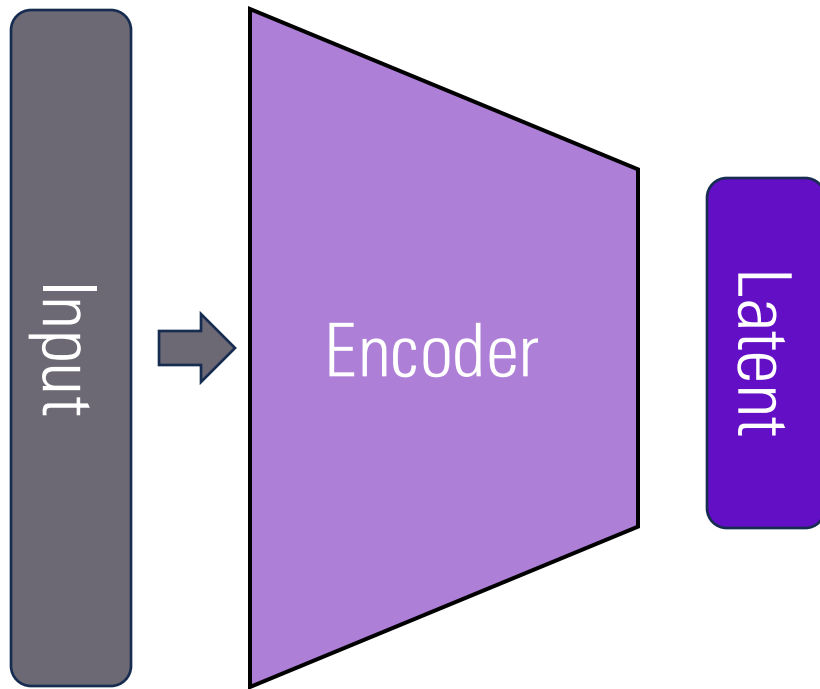


Extract meaningful, information-dense features from the data

<https://huggingface.co/spaces/mteb/leaderboard>



# A unified “API” for complex data



Different data modalities (images, text, etc) all map to embeddings - simple vectors of numbers.