

# Compressing Latent Space via Least Volume

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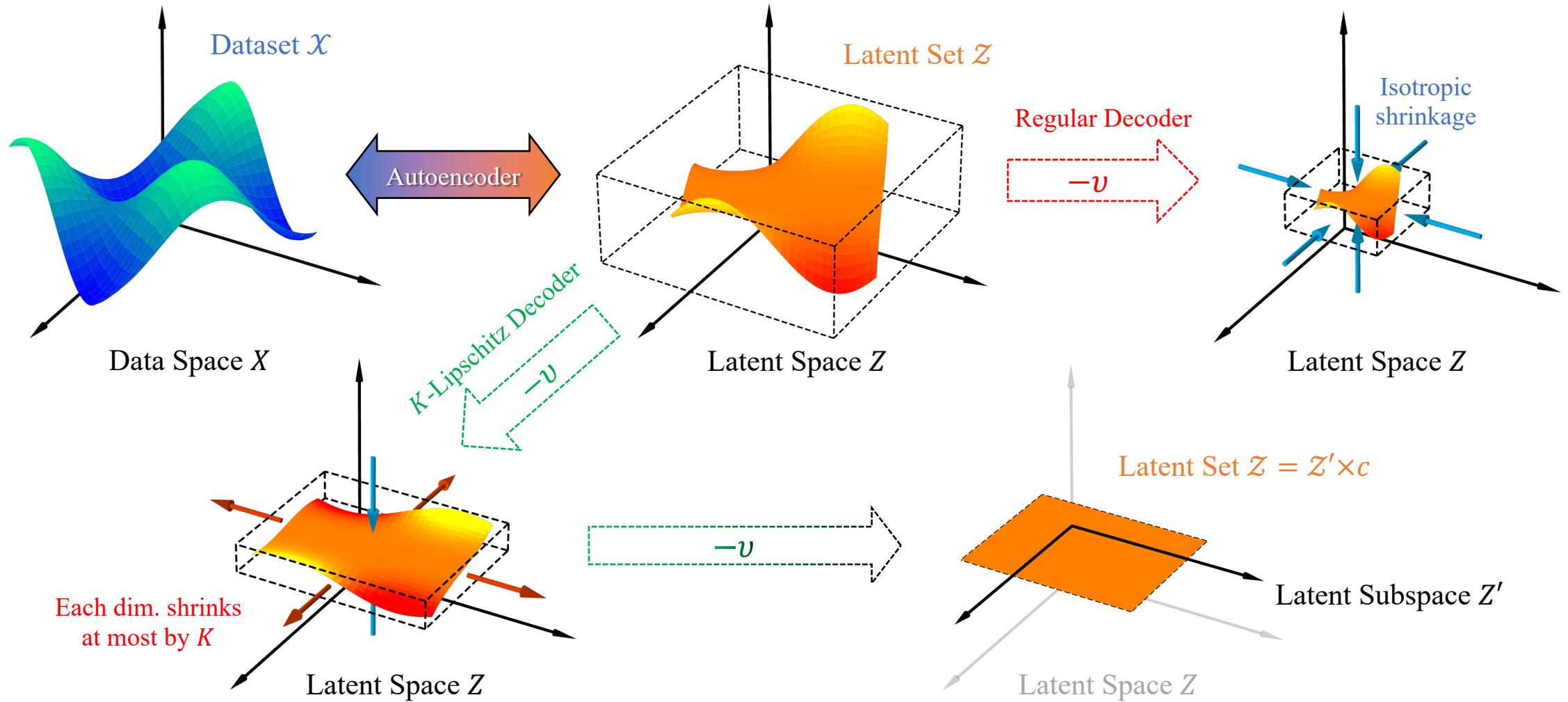
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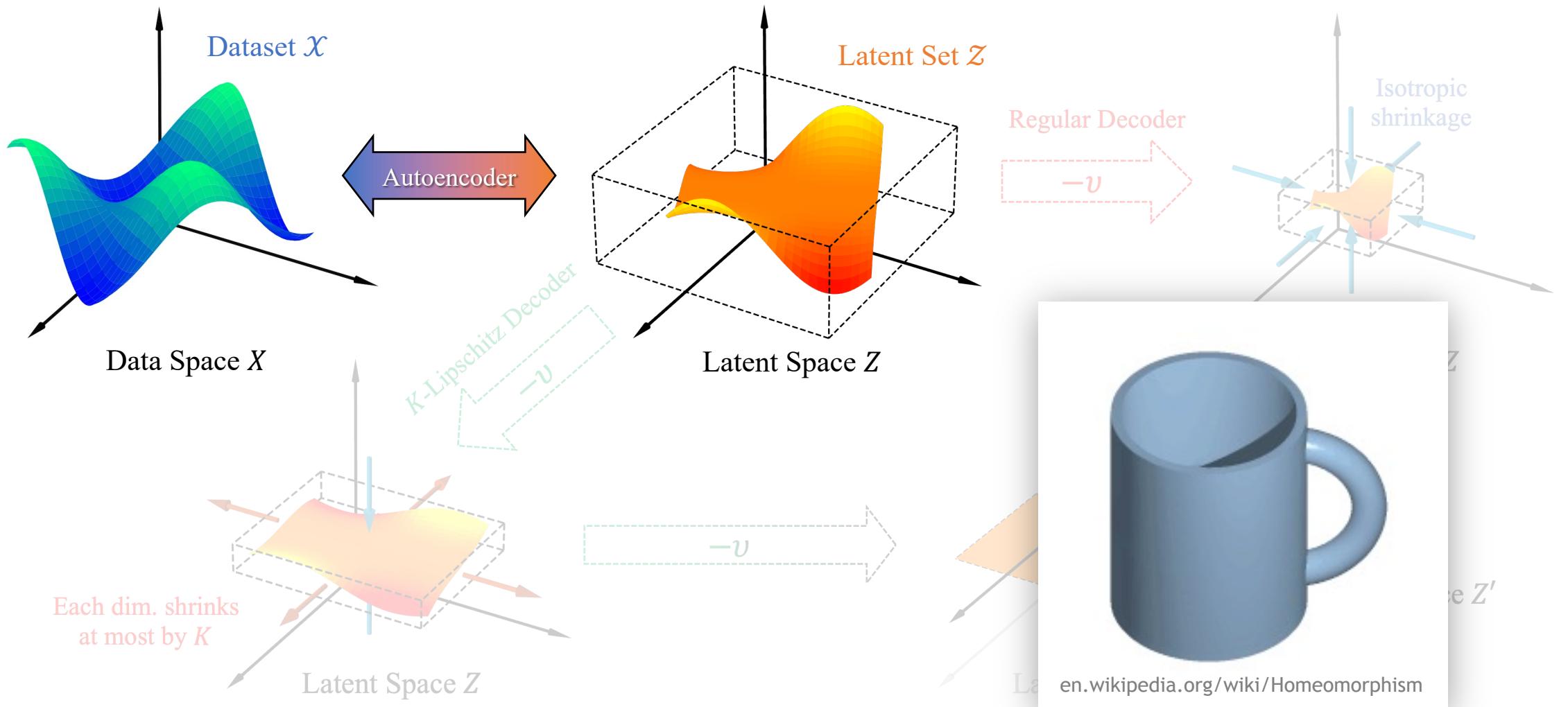
# Motivation

- Real-world high dimensional datasets may live on low dimensional nonlinear manifolds.
- Retrieving the least dimensional representations of the datasets could be beneficial:
  - It may alleviate the curse of dimensionality the most.
  - It may improve some downstream tasks, e.g., data generation.
  - It may let us study the dataset's structure more easily.

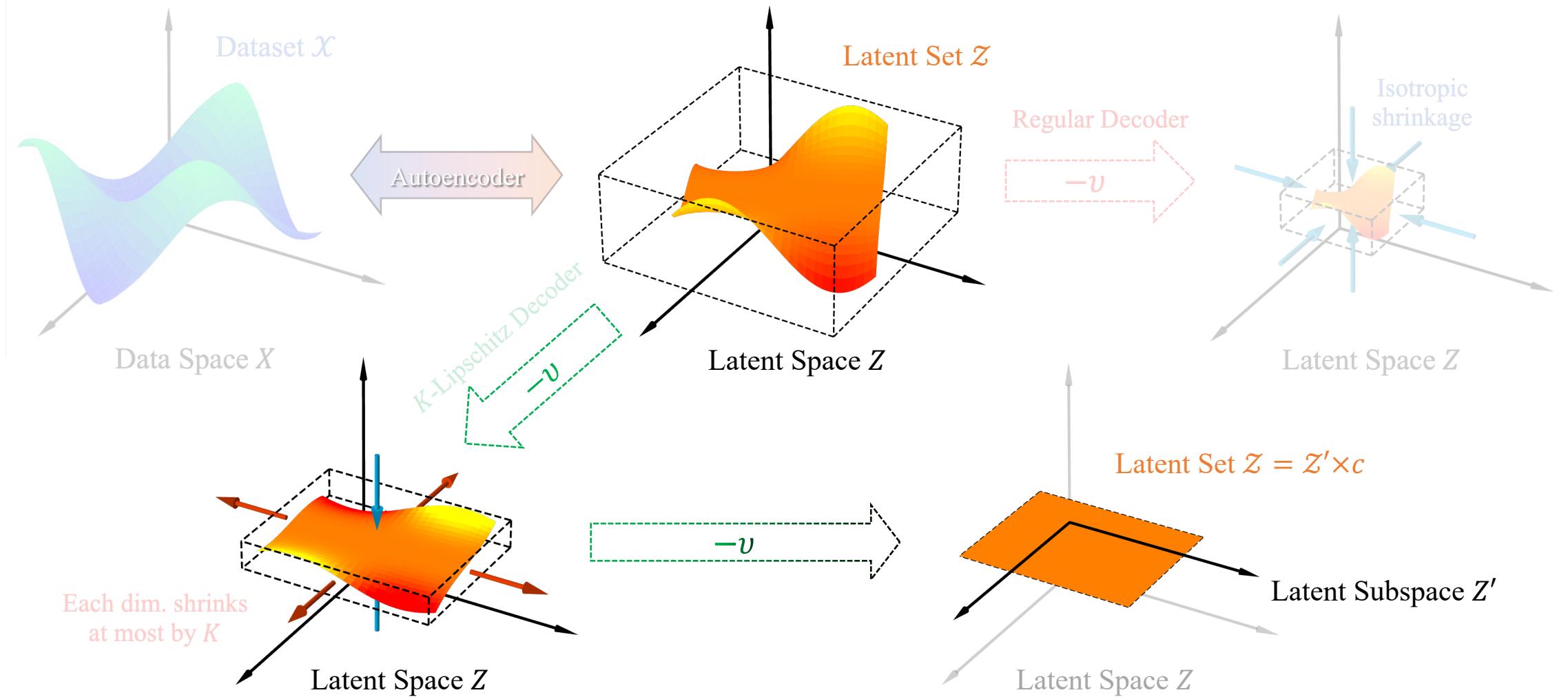
**Goal:** Obtain the least dimensional representation automatically.

# Methodology: Least Volume

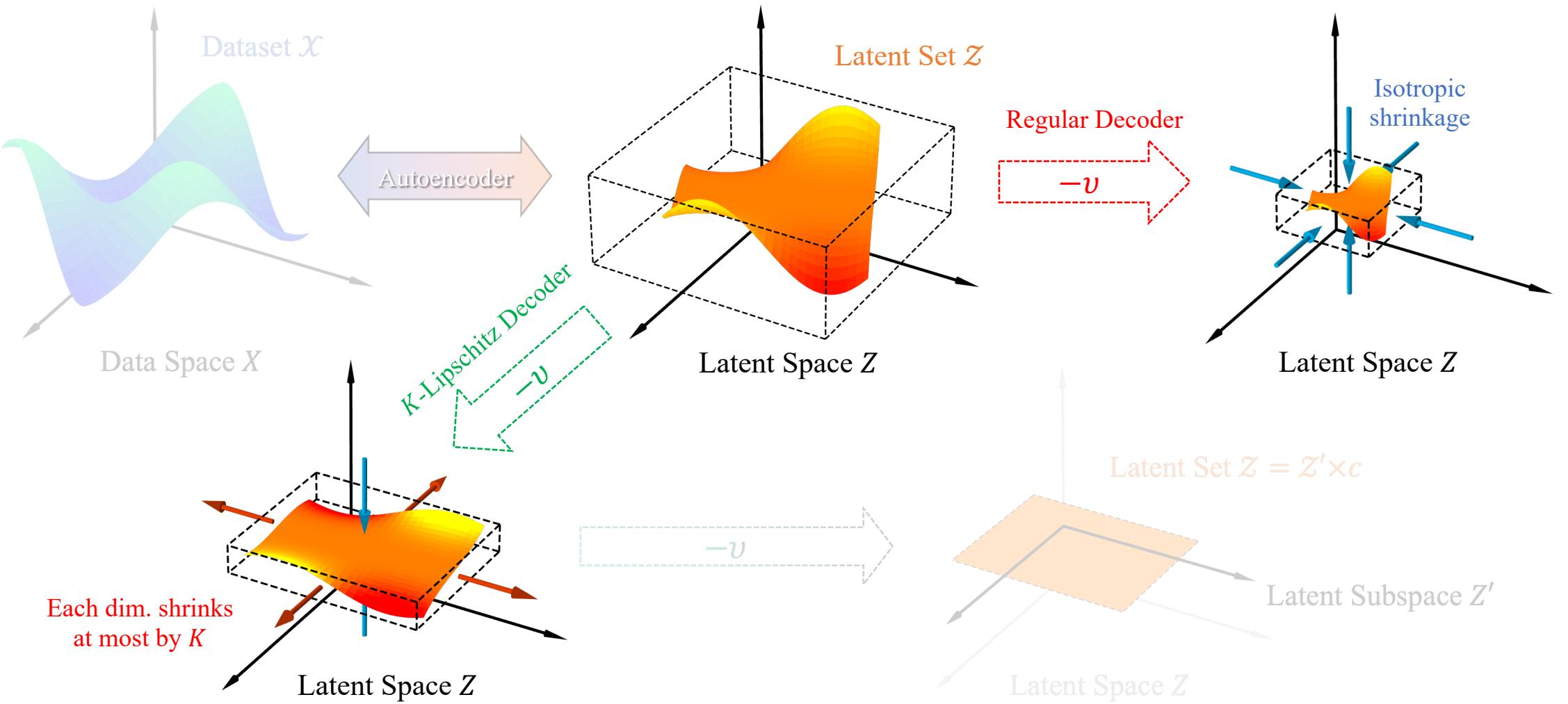




A *continuous* autoencoder's latent set  $Z$  is *homeomorphic* to the dataset  $\mathcal{X}$ .



A flattened latent set consumes less space  $\nu = \prod_i \sigma_i$  than a curved one.



A  $K$ -Lipschitz decoder prevents the latent set's isotropic shrinkage.

# Least Volume Formulation

1. Make the decoder  $g$   $K$ -Lipschitz with *spectral normalization*.
2. Minimize the loss function with the volume penalty:

$$L = J + \lambda L_{\text{vol}}$$

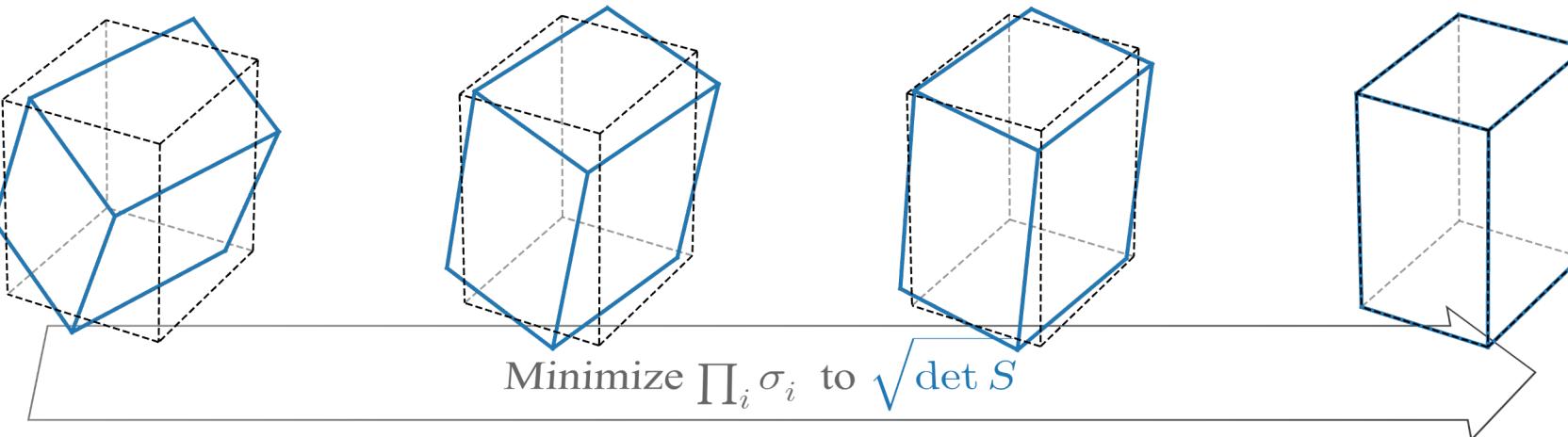
Reconstruction Loss:  $J = \mathbb{E}\|g \circ e(x) - x\|$

Volume Penalty:  $L_{\text{vol}} = \sqrt[m]{\prod_i (\sigma_i + \eta)}$   $\leftarrow \dots \nu = \prod_i \sigma_i$

We can then reduce the autoencoder's latent dimension *automatically*.

# PCA is a Linear Special Case of Least Volume

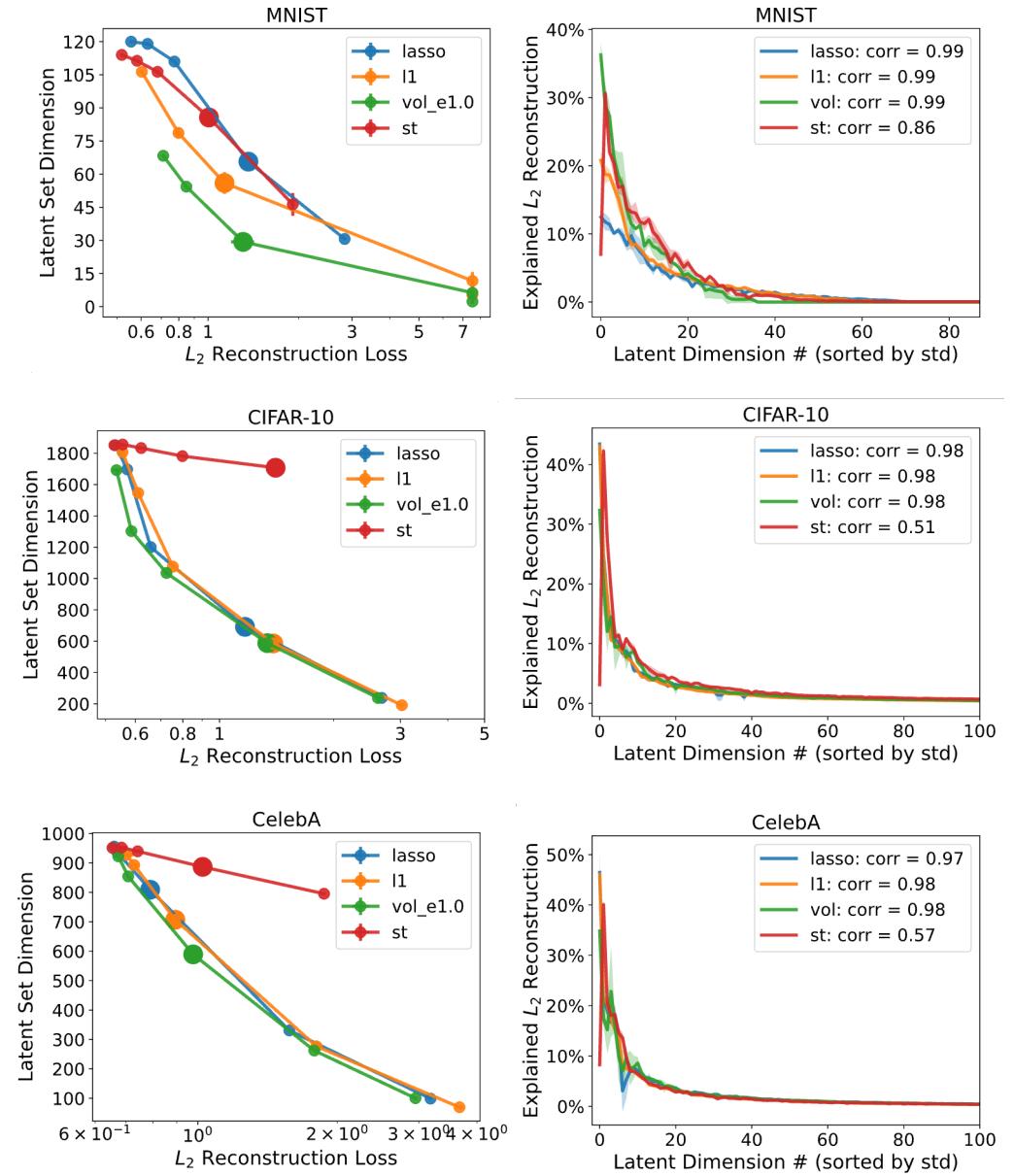
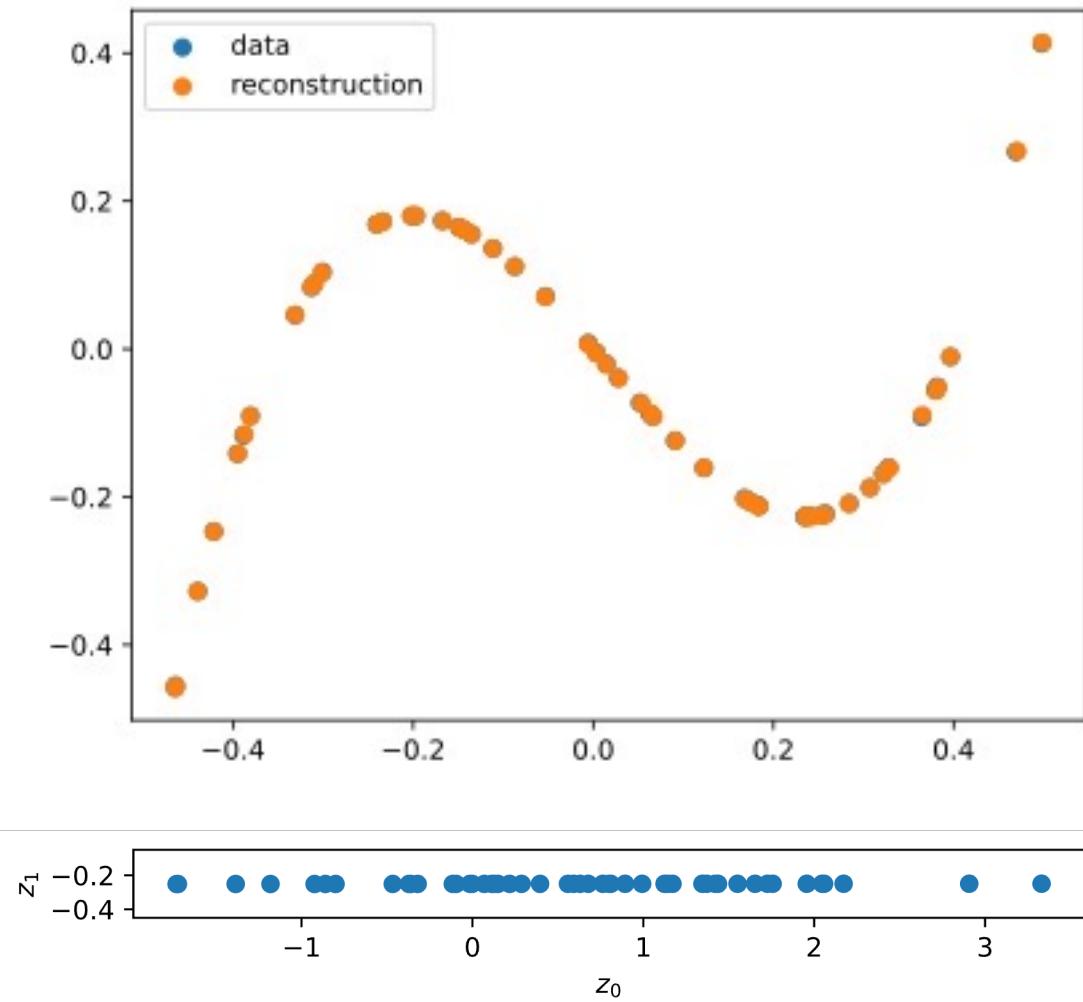
- $\nu = \prod_i \sigma_i$  is a tight upper-bound of  $\sqrt{\det S} = \sqrt{\det \text{Cov}(\mathcal{Z})}$ .



- If the autoencoder is linear, then least volume turns it into PCA.

*Nonlinear autoencoders may have a similar ordering effect.*

# Experiment Results

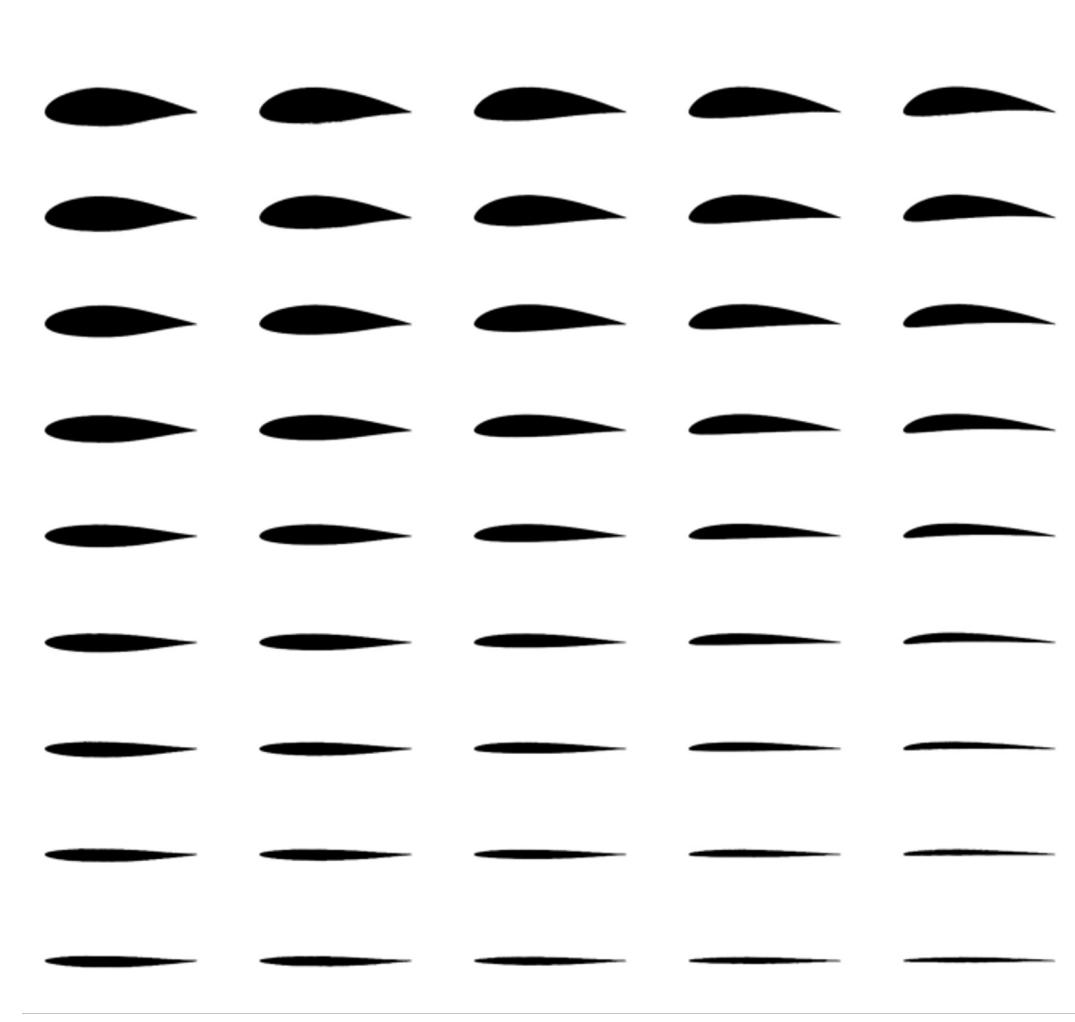
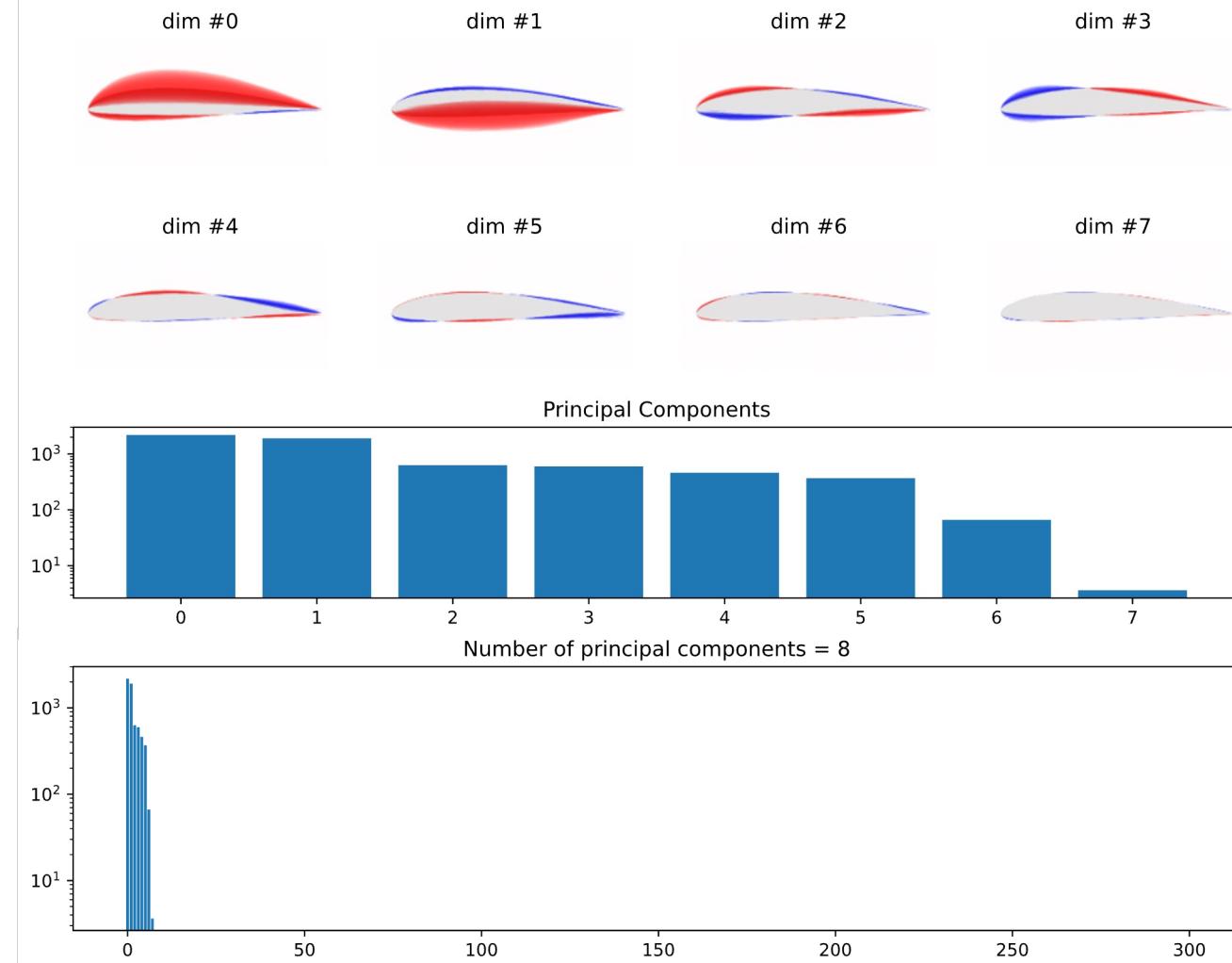


Motivation > Methodology > Properties > **Results** > Applications



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# Application: Design Analysis & Parameterization

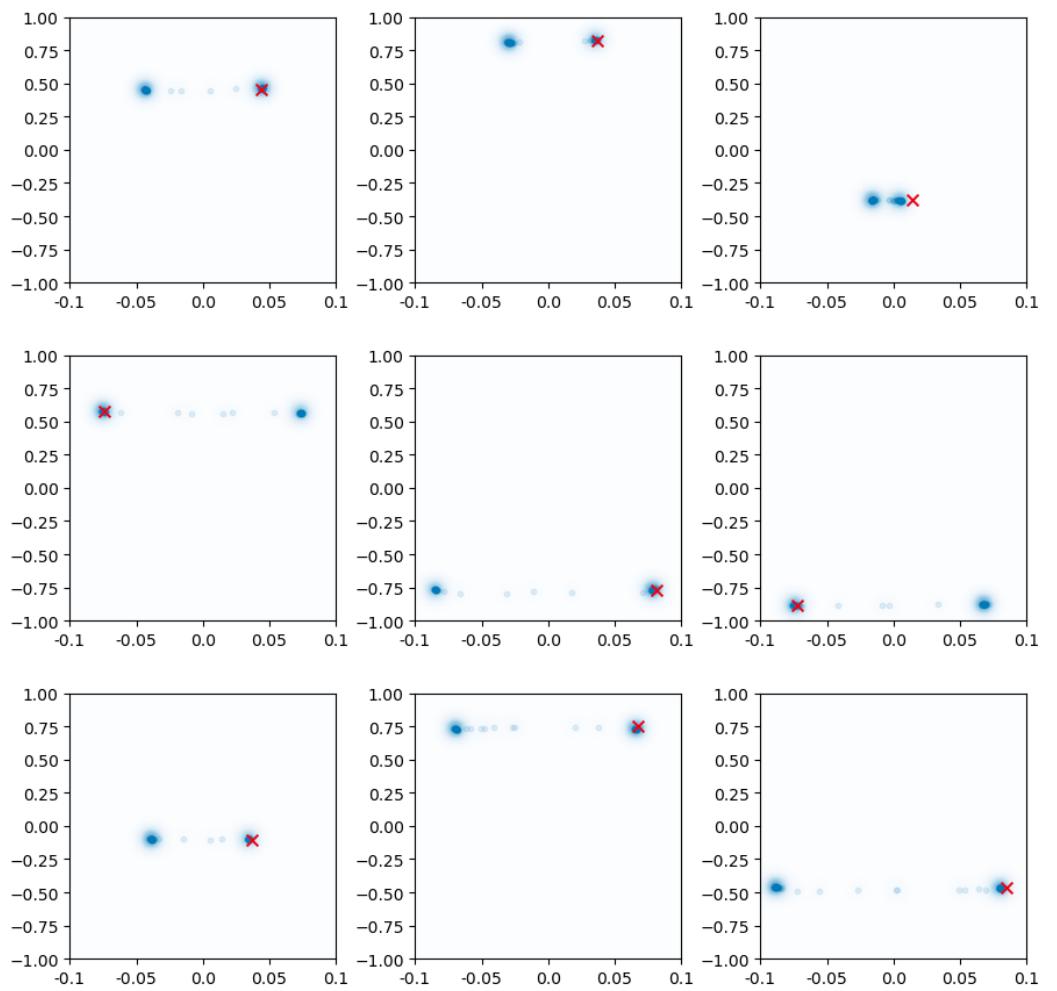
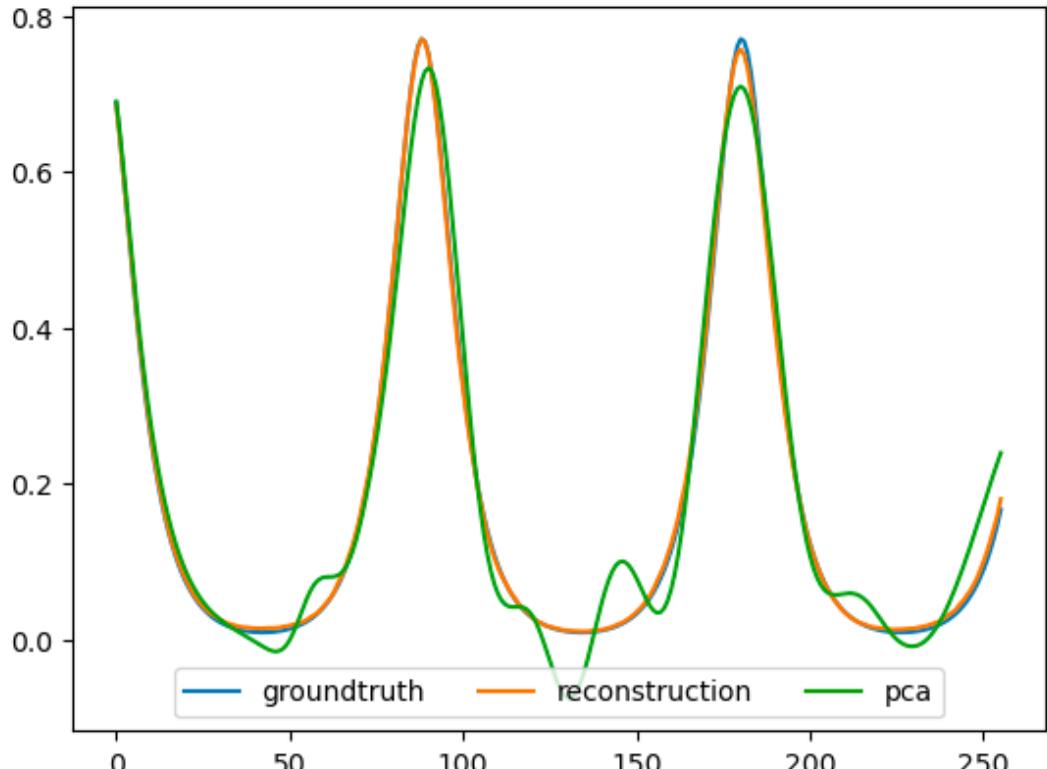


Motivation > Methodology > Properties > Results > **Applications**



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# Application: Time Series Analysis & System Identification



Motivation > Methodology > Properties > Results > Applications



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# Thanks for watching!

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GitHub: [IDEALLab/Least\\_Volume\\_ICLR2024](https://github.com/IDEALLab/Least_Volume_ICLR2024)



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