

# Quiz: Dimensionality Reduction

Lecture series „Machine Learning“

Niels Landwehr

Research Group „Data Science“  
Institute of Computer Science  
University of Hildesheim

# Quiz: Intrinsic Dimensionality of Data

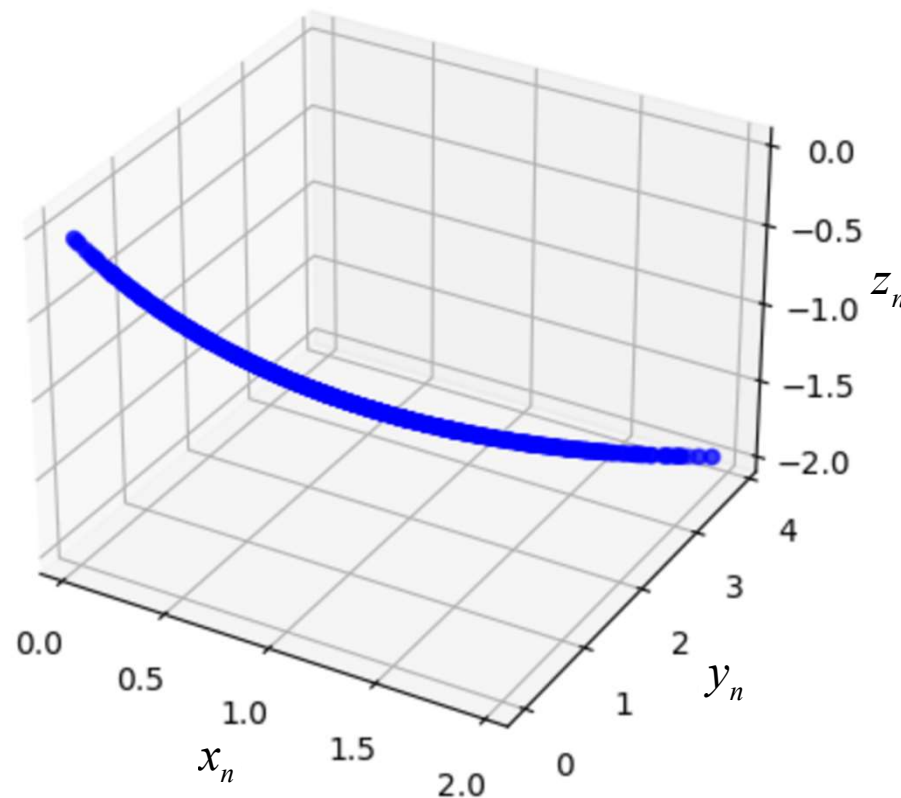
- Assume a data set of 1000 points  $\{\mathbf{x}_1, \dots, \mathbf{x}_{1000}\}$  with  $\mathbf{x}_n = (x_n, y_n, z_n) \in \mathbb{R}^3$  generated as follows:
  - Draw  $a_n \sim \text{uniform}(0,1)$ ,  $b_n \sim \text{uniform}(0,1)$
  - Set  $x_n = a_n + b_n$ ,  $y_n = a_n^2 + b_n^2 + 2a_nb_n$ ,  $z_n = -a_n - b_n$
- **Question:** What is the resulting intrinsic dimensionality of the data? In other words, which embedding dimension would a sufficiently large autoencoder model need to perfectly fit the data?
  - The intrinsic dimensionality of the data is 3, so three-dimensional embedding is needed
  - The intrinsic dimensionality of the data is 2, so a two-dimensional embedding is needed
  - The intrinsic dimensionality of the data is 1, so a one-dimensional is enough
  - Cannot say based on the data provided

# Solution: Intrinsic Dimensionality of Data

- **Solution:** The intrinsic dimensionality of the data is one, so a one-dimensional embedding is enough
- We can rewrite the data generating process as follows:
  - Draw  $a_n \sim \text{uniform}(0,1)$ ,  $b_n \sim \text{uniform}(0,1)$
  - Set  $c_n = a_n + b_n$
  - Set  $x_n = c_n$ ,  $y_n = c_n^2$ ,  $z_n = -c_n$
- Now it is clear that the data is one-dimensional, because running through all possible values of  $c_n$  will generate all possible data points

# Solution: Intrinsic Dimensionality of Data

- **Solution:** The intrinsic dimensionality of the data is one, so a one-dimensional embedding is enough
- Here is a plot of the data:



# Quiz: PCA

- Assume the same data set  $\{\mathbf{x}_1, \dots, \mathbf{x}_{1000}\}$  as above:
  - Draw  $a_n \sim \text{uniform}(0,1), b_n \sim \text{uniform}(0,1)$
  - Set  $x_n = a_n + b_n, y_n = a_n^2 + b_n^2 + 2a_nb_n, z_n = -a_n - b_n$
- **Question:** what is the needed number of components  $K$  in a PCA so that the PCA can perfectly reconstruct the data (without reconstruction error)?
  - $K=3$  is needed, data does not lie in any linear subspace (except the original space)
  - $K=2$  is needed, data lies on a two-dimensional linear subspace
  - $K=1$  is needed, data lies on a one-dimensional linear subspace

# Solution: PCA

- **Solution:**  $K=2$  is needed, data lies on a two-dimensional linear subspace
- As stated above, the data has one intrinsic dimension  $c$ , and all three coordinates in the original space can be computed from that intrinsic dimension:

$$x_n = c_n \quad y_n = c_n^2 \quad z_n = -c_n$$

- The first coordinate  $x_n = c_n$  and the last coordinate  $z_n = -c_n$  are linear functions of the intrinsic dimension  $c$ , so they lie within a one-dimensional linear subspace
- The second coordinate  $y_n = c_n^2$  however depends non-linearly on  $c$ , thus one more dimension in the PCA for this coordinate is needed

