report.md 2024-12-02

Analysis of Model Performance

Reconstruction Errors

Linear Autoencoder

Reconstruction Error: 0.4096

Nonlinear Autoencoder

• Reconstruction Error: **0.4910**

• PCA (Principal Component Analysis)

• Reconstruction Error: 0.4459

Comparison

Linear Autoencoder

- The overlap between original data (blue points) and reconstructed data (red points) is substantial.
- The model retains linear data structure effectively, with the lowest reconstruction error of **0.4096**.

Nonlinear Autoencoder

- Some reconstructed points deviate significantly. This may be due to the complexity of the model or the fact that the data itself is relatively linear, limiting the nonlinear model's strengths.
- Reconstruction error is slightly higher at **0.4910**.
- The loss curve shows that the loss only drops significantly at very large epochs. This indicates that a high number of epochs is necessary for the model to converge effectively. If the number of epochs is insufficient, the model performs poorly.

PCA

- The overlap between original and reconstructed data is moderate.
- Reconstruction performance is less accurate than the Linear Autoencoder but better than the Nonlinear Autoencoder.
- Reconstruction error is **0.4459**, as PCA is limited to linear transformations.

Conclusion

1. Linear Autoencoder:

 Lowest reconstruction error, making it suitable for tasks requiring high fidelity in data reconstruction.

2. Nonlinear Autoencoder:

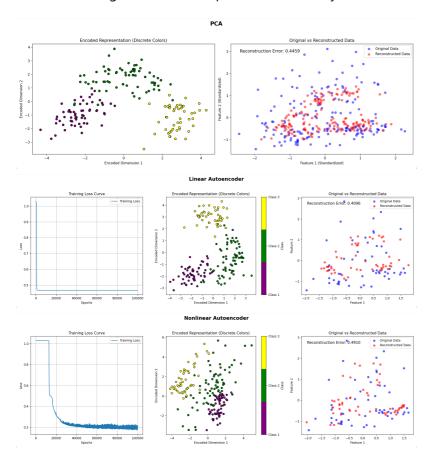
- While its reconstruction error is the highest, it captures nonlinear features better. However, its class separation is the weakest, with significant overlap among different classes.
- Recommended for complex or nonlinear data distributions, such as clustering or classification tasks.

report.md 2024-12-02

3. **PCA**:

- Computationally efficient and provides moderate reconstruction quality.
- Both PCA and Linear Autoencoder produce more concentrated reconstructed data points compared to the Nonlinear Autoencoder.

• A good choice for quick dimensionality reduction in linearly distributed data.



Github

View the code on GitHub