

UMAP: Uniform Manifold Approximation and Projection

Leland McInnes 1 , John Healy 1 , Nathaniel Saul 2 , and Lukas Großberger $^{3,\;4}$

1 Tutte Institute for Mathematics and Computing 2 Department of Mathematics and Statistics, Washington State University 3 Ernst Strüngmann Institute for Neuroscience in cooperation with Max Planck Society 4 Donders Institute for Brain, Cognition and Behaviour, Radboud Universiteit

DOI: 10.21105/joss.00861

Software

■ Review 🗗

■ Repository 🖸

Archive □

Submitted: 19 July 2018 Published: 02 September 2018

License

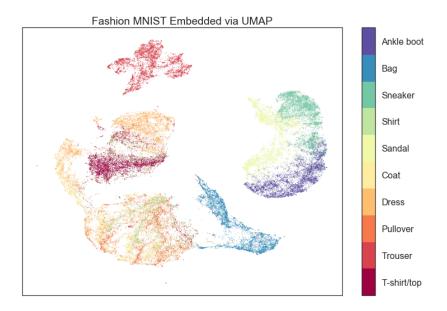
Authors of papers retain copyright and release the work under a Creative Commons Attribution 4.0 International License (CC-BY).

Summary

Uniform Manifold Approximation and Projection (UMAP) is a dimension reduction technique that can be used for visualisation similarly to t-SNE, but also for general non-linear dimension reduction. UMAP has a rigorous mathematical foundation, but is simple to use, with a scikit-learn compatible API. UMAP is among the fastest manifold learning implementations available – significantly faster than most t-SNE implementations.

UMAP supports a number of useful features, including the ability to use labels (or partial labels) for supervised (or semi-supervised) dimension reduction, and the ability to transform new unseen data into a pretrained embedding space.

For details of the mathematical underpinnings see (McInnes & Healy, 2018). The implementation can be found at (McInnes, Healy, Saul, & Grossberger, 2018).



1



References

McInnes, L., & Healy, J. (2018). UMAP: Uniform Manifold Approximation and Projection for Dimension Reduction. $ArXiv\ e\text{-}prints.$

McInnes, L., Healy, J., Saul, N., & Grossberger, L. (2018). UMAP. Retrieved July 22, 2018, from https://github.com/lmcinnes/umap

2