

dml: Distance Metric Learning in R

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DOI: 10.21105/joss.01036

Software

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Submitted: 18 October 2018 Published: 23 October 2018

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Summary

Distance metric is widely used in the machine learning literature. We used to choose a distance metric according to a priori (e.g. Euclidean Distance, L1 Distance, etc.) or according to the result of cross validation within small class of functions (e.g. choosing order of polynomial for a kernel). Actually, with priori knowledge of the data, we could learn a more suitable distance metric with (semi-)supervised distance metric learning techniques. dml (Tang, Gao, & Xiao, 2015) is such an R package aims to implement a collection of algorithms for (semi-)supervised distance metric learning.

The dml package provides native R implementations for a collection of Distance Metric Learning algorithms, including both global and local methods such as Relevant Component Analysis (Shental, Hertz, Weinshall, & Pavel, 2002), Discriminative Component Analysis (Peltonen, Goldberger, & Kaski, 2007), and Local Fisher Discriminant Analysis (Sugiyama, 2006). A list of all the implemented algorithms can be found in the dml package reference manual. These methods are widely applied in feature extraction, dimensionality reduction, clustering, information retrieval, and computer vision problems.

Additionally, implementations for the variants of the methods are also available in dml package. For example, since it was built on top of the lfda (Tang, 2017; Tang & Li, 2016) package, users also have access to the family of Local Fisher Discriminant Analysis methods, which includes Local Fisher Discriminant Analysis, Kernel Local Fisher Discriminant Analysis, and Semi-supervised Local Fisher Discriminant Analysis (Sugiyama, Idé, Nakajima, & Sese, 2010). To make the results of these methods easy for users to interprete and analyze, both static and interactive visualizations for the results are available through ggfortify (Horikoshi & Tang, 2018; Tang, Horikoshi, & Li, 2016) and autoplotly (Tang, 2018a, 2018b) packages respectively.

References

Horikoshi, M., & Tang, Y. (2018). *Ggfortify: Data visualization tools for statistical analysis results*. Retrieved from https://CRAN.R-project.org/package=ggfortify

Peltonen, J., Goldberger, J., & Kaski, S. (2007). Fast semi-supervised discriminative component analysis. In *Machine learning for signal processing*, 2007 ieee workshop on (pp. 312–317). IEEE. doi:10.1109/MLSP.2007.4414325

Shental, N., Hertz, T., Weinshall, D., & Pavel, M. (2002). Adjustment learning and relevant component analysis. In *European conference on computer vision* (pp. 776–790). Springer.



Sugiyama, M. (2006). Local fisher discriminant analysis for supervised dimensionality reduction. In *Proceedings of the 23rd international conference on machine learning* (pp. 905–912). ACM. doi:10.1145/1143844.1143958

Sugiyama, M., Idé, T., Nakajima, S., & Sese, J. (2010). Semi-supervised local fisher discriminant analysis for dimensionality reduction. *Machine learning*, 78(1-2), 35. doi:10.1007/s10994-009-5125-7

Tang, Y. (2017). Local fisher discriminant analysis. Retrieved from https://CRAN. R-project.org/package=lfda

Tang, Y. (2018a). Autoplotly: An R package for automatic generation of interactive visualizations for statistical results. Retrieved from https://CRAN.R-project.org/package=autoplotly

Tang, Y. (2018b). Autoplotly: An R package for automatic generation of interactive visualizations for statistical results. *Journal of Open Source Software*, 3(24), 657. doi:10.21105/joss.00657

Tang, Y., & Li, W. (2016). Lfda: An R package for local fisher discriminant analysis and visualization. arXiv preprint arXiv:1612.09219. Retrieved from https://arxiv.org/abs/1612.09219

Tang, Y., Gao, T., & Xiao, N. (2015). *Dml: Distance metric learning in R.* Retrieved from https://CRAN.R-project.org/package=dml

Tang, Y., Horikoshi, M., & Li, W. (2016). Ggfortify: Unified interface to visualize statistical result of popular R packages. *The R Journal*, 8(2), 478–489. Retrieved from https://journal.r-project.org/