

dml: Distance Metric Learning in R

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Software

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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 (Y. Tang, Gao, & Xiao, 2015) is such an R package aims to implement the state-of-the-art algorithms for (semi-)supervised distance metric learning.

The dml package provides native R implementations of state-of-the-art algorithms for Distance Metric Learning, including both global and local methods such as Relevant Component Analysis, Discriminative Component Analysis, and Local Fisher Discriminant Analysis (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, it builds on top of the lfda (Y. Tang, 2017; Tang, Yuan, Li, & Wenxuan, 2016) package to provide access to Local Fisher Discriminant Analysis family of methods, which includes Local Fisher Discriminant Analysis, Kernel Local Fisher Discriminant Analysis, and Semi-supervised Local Fisher Discriminant Analysis. 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 the ggfortify (Horikoshi & Tang, 2018; Y. Tang, Horikoshi, & Li, 2016) and the autoplotly (Y. Tang, 2018a, 2018b) package respectively.

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