

## Quantifying variable importance in cluster analysis

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The quantification of variable importance in cluster analysis is of interest in order to interpret and understand the impact of the variables on a clustering, and potentially also for variable selection. For general clustering methods, it can be measured by comparing a clustering with all variables with a clustering in which a variable has been left out or permuted ([1, 2]). We compare these two approaches regarding their ability to tell apart meaningful from noise variables.

A potential concern regarding clustering mixed continuous/categorical variables is that certain methods may be unduly dominated by either the continuous or the categorical variables ([3]). We address this by comparing methods such as latent class model-based clustering, distance-based clustering using Gower's distance with various weighting/standardisation schemes, or KAMILA regarding the relative importance of the continuous and categorical variables using a comprehensive simulation study and real data.

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