Choosing the Best Method for Gene Expression log- log Linear Models Using Multiple CART Trees

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Microarray and RNA-Seq techniques are used to infer genes showing differential expressions on treatment conditions through the analysis of loglog linear models for the expression with treatment compared with control condition. Due to costs and technical limitations usually the experiments present small-sized samples and high contamination; therefore, choosing the estimation method for coefficients of such models becomes a challenge [1]. Herein, we simulate microarray and RNA-Seq experiments and analyze a log-log linear model with contaminations at both conditions, varying key features: the sample size n, contamination type (*light-tailed* or *heavy-tailed*), contamination proportion p, and error variance σ^2 . For each features configuration we computed the accuracy at each method among least absolute deviations (l_1) , ordinary least squares (l_2) , and Huber M-Estimators (HM). Using this information, we built a machine learning that, based on classification CART trees [2], automatically decides the best method depending on simple questions. Restricted to light tails, l_2 leads if n is small and σ^2 low, while l_1 leads if σ^2 is moderate and p high, and for p low and σ^2 moderate HM leads. In case of heavy tails, HM leads if p is moderate. Simulation results on method decisions agree with theoretical analysis [3], adding more information for non-extreme conditions, and show good sensitivity and specificity in true experiments.

References

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