

Improving MLE via a Non-extensive Information Measure

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Abstract

Although the maximum likelihood estimator enjoys asymptotic optimality properties, its finite-sample performance for a small or moderate sample size can be much improved when the Log-Likelihood is replaced by a L_q -Likelihood, which is motivated from a non-extensive measure of information (in contrast to the additive Kullback-Leiber information).

The properties of the resulting estimator, ML_qE , are studied via asymptotic analysis and computer simulations. The behavior of the ML_qE is characterized by the degree of distortion q applied to the assumed model to amplify or diminish the density value. When q is properly chosen for small and moderate sample sizes, the ML_qE successfully trades bias for precision, resulting in a substantial reduction of the mean squared error. When the sample size is large and q tends to 1, a necessary and sufficient condition to ensure a proper asymptotic normality and efficiency of ML_qE is established. The advantage of the new estimation method is more clearly seen for higher dimensional estimations. The talk is based on joint work with Davide Ferrari.