

Near-consistent robust estimations of moments for unimodal distributions

Tuban Lee

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1 **Descriptive statistics for parametric models currently rely heavily**
2 **on the accuracy of distributional assumptions. Here, leveraging the**
3 **structures of parametric distributions and their central moment kernel**
4 **distributions, a class of estimators, consistent simultaneously for both**
5 **a semiparametric distribution and a distinct parametric distribution, is**
6 **proposed. These efficient estimators are robust to both gross errors**
7 **and departures from parametric assumptions, making them ideal**
8 **for estimating the mean and central moments of common unimodal**
9 **distributions. This article also illuminates the understanding of the**
10 **common nature of probability distributions and the measures of them.**

A. Invariant Moments. Most robust location estimators commonly used are symmetric owing to the prevalence of symmetric distributions. A γ -weighted Hodges-Lehmann mean ($\text{WHLM}_{k,\epsilon,\gamma,n}$) can achieve consistency for any γ -symmetric distribution. However, it falls considerably short of effectively handling a broad spectrum of other common distributions. Shifting from semiparametrics to parametrics, consider an estimator with a non-sample-dependent breakdown point (defined in Subsection ??) that is consistent simultaneously for both a semiparametric class of distributions and a distinct parametric distribution, such a robust estimator is named with the prefix ‘invariant’ followed by the population parameter it is consistent with. Here, the recombined I -statistic is defined as

$$\text{RI}_{d,\mathbf{k}_1,\mathbf{k}_2,k_1,k_2,\epsilon_1,\epsilon_2,\gamma_1,\gamma_2,n,LU_1,LU_2} := \lim_{c \rightarrow \infty} \left(\frac{(LU_1 \mathbf{k}_1, k_1, \epsilon_1, \gamma_1, n + c)^{d+1}}{(LU_2 \mathbf{k}_2, k_2, \epsilon_2, \gamma_2, n + c)^d} - c \right),$$