

ON THE BIAS OF THE JACKKNIFE ESTIMATE OF VARIANCE¹

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Using machinery developed earlier for the covariances of symmetric statistics, we consider various aspects of the bias of the jackknife estimate of variance.

1. Introduction

The jackknife estimate of variance (Quenouille (1949, 1956), Tukey (1958)) can be described as follows. Given a symmetric function h of iid arguments X_1, X_2, \dots, X_m , it is desired to estimate $\sigma^2 = \text{Var } h$. With an augmented supply X_1, X_2, \dots, X_n where $n = m+1$, or more generally $n \geq m+1$, one forms $Q = \binom{n-1}{m}^{-1} \sum_{|I|=m} [h(X_I) - \bar{h}]^2$ where $\bar{h} = \binom{n}{m}^{-1} \sum_{|I|=m} h(X_I)$ and $X_I \equiv (X_{i_1}, X_{i_2}, \dots, X_{i_m})$ with $I = \{i_1, i_2, \dots, i_m\}$. Several papers (Efron and Stein (1981), Karlin and Rinott (1982), Bhargava (1983), Vitale (1984), Steele (1986)) have considered the bias relation

$$(1.1) \quad \sigma^2 \leq EQ,$$

which has come to be known as the Efron–Stein inequality. Our purpose here is to investigate aspects of (1.1) including (i) an alternate proof with variant forms of the condition for equality, (ii) a sharpening, (iii) a complementary upper bound, and (iv) a consideration of Q as an estimator which is “contaminated” by estimators of other parameters.

2. Preliminaries

If X_1, X_2, \dots, X_n are iid random variables and h is a symmetric function of m of them with $Eh^2 < \infty$, then we set

$$r_k = \text{Cov}[h(X_I), h(X_J)]$$

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