Density Estimation with Replicate Heteroscedastic Measurements

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Received: 29 January 2008 / Revised: 4 September 2008 / Published online: 26 March 2009 © The Institute of Statistical Mathematics, Tokyo 2009

Abstract We present a deconvolution estimator for the density function of a random variable from a set of independent replicate measurements. We assume that measurements are made with normally distributed errors having unknown and possibly heterogeneous variances. The estimator generalizes well-known deconvoluting kernel density estimators, with error variances estimated from the replicate observations. We derive expressions for the integrated mean squared error and examine its rate of convergence as $n \to \infty$ and the number of replicates is fixed. We investigate the finite-sample performance of the estimator through a simulation study and an application to real data.

Keywords Bandwidth \cdot Bootstrap \cdot Deconvolution \cdot Hypergeometric series \cdot Measurement error