

MINIMUM  $L_1$ -PENALIZED DISTANCE ESTIMATORS OF A DENSITY AND  
ITS DERIVATIVES

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*Abstract:* Let  $F$  be an  $(m + 1)$ -times differentiable distribution function (df) generating the data. Let  $f$  be the density of  $F$ . Let  $F_n$  denote the empirical df. The paper concerns fitting an  $(m + 1)$ -times differentiable function  $G$  to the data by minimizing  $d_n(G) = \|F_n - G\|_1 + \beta(n)\|G^{(m+1)}\|_1$ , where  $\|\cdot\|_p$ ,  $p \geq 1$ , denotes the  $L_p$ -norm and  $\beta(n) > 0$  is a sequence of smoothing parameters. Let  $\hat{F}_n$  be an (approximate) minimizer of the above problem. We establish an upper bound for  $E\|\hat{F}_n^{(i)} - F^{(i)}\|_1$ ,  $i = 1, \dots, m$ , with respect to the choice of  $\beta$ . In particular, the choice of  $\beta \sim n^{-1/(m+1)}$  results in the optimal  $L_1$ -rate of convergence of  $\hat{F}_n'$  to  $f$ . The estimation  $E\|\hat{F}_n^{(i)} - F^{(i)}\|_2^2$  is also evaluated.

**2000 AMS Mathematics Subject Classification:** Primary: -; Secondary: -;

**Key words and phrases:** -

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