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## Density estimation of a mixture distribution with unknown point-mass and normal error



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### ABSTRACT

We consider the model  $Y=X+\xi$  where Y is observable,  $\xi$  is a noise random variable with density  $f_{\xi}$ , X has an unknown mixed density such that  $\mathbb{P}(X=X_c)=1-p$ ,  $\mathbb{P}(X=a)=p$  with  $X_c$  being continuous and  $p\in(0,1)$ ,  $a\in\mathbb{R}$ . Typically, in the last decade, the model has been widely considered in a number of papers for the case of fully known quantities  $a,f_{\xi}$ . In this paper, we relax the assumptions and consider the parametric error  $\xi\sim\sigma N(0,1)$  with an unknown  $\sigma>0$ . From i.i.d. copies  $Y_1,\ldots,Y_m$  of Y we will estimate  $(\sigma,p,a,f_{X_c})$  where  $f_{X_c}$  is the density of  $X_c$ . We also find the lower bound of convergence rate and verify the minimax property of established estimators. © 2021 Published by Elsevier B.V.

### 1. Introduction

In applications, we often have the problem of finding the probability distribution of a random variable X from measurements contaminated with a noise  $\xi$ . Simplicity, we can assume the additive model  $Y = X + \xi$  with Y observed. Estimation of the target density  $f_X$  of X from observations of the random variable Y is called the (statistical) deconvolution problem. The literature of the specific problem has grown very rapidly in last decades (see Devroye, 1989; Meister, 2009 and references therein). Precisely, in the present paper, we assume the data model

$$Y_i = X_i + \xi_i, \quad j = \overline{1, m}, \tag{1}$$

where the  $Y_j$ 's are i.i.d. observable copies of the random variable Y, the  $X_j$ 's and the  $\xi_j$ 's,  $j = \overline{1, m}$ , are mutually independent and have the same distribution as X,  $\xi$  respectively. As is well known, any estimation tool of deconvolution problem is based on a combination of presumptions about the target density function and the density function of error.

We first discuss shortly the presumption on the target density function. In the deconvolution literature, most of the research is focused on continuous distributions. However, in some applications, the target random variable *X* is not

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