

# Moderate deviations and law of the iterated logarithm in $L_1(\mathbb{R}^d)$ for kernel density estimators

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## Abstract

Let  $f_n(x)$  be the non-parametric kernel density estimator of a density function  $f(x)$  based on a kernel function  $K$ . In this paper, we first prove two moderate deviation theorems in  $L_1(\mathbb{R}^d)$  for  $\{f_n(x), n \geq 1\}$ . Then, as an application of the moderate deviations, we obtain a law of the iterated logarithm for  $\{\|f_n - Ef_n\|_1, n \geq 1\}$ .

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## 1. Introduction and main results

Let  $\{X_i; i \geq 1\}$  be a sequence of independent and identically distributed (*i.i.d.*) random variables taking values in  $\mathbb{R}^d$ , defined on a probability space  $(\Omega, \mathcal{F}, P)$  with unknown density function  $f(x)$ . Let  $K$  be a measurable function such that

$$K \geq 0, \quad \int_{\mathbb{R}^d} K(x) dx = 1. \quad (1.1)$$

The kernel density estimator of  $f$ , based on the kernel function  $K$ , is defined by

$$f_n(x) = \frac{1}{na_n^d} \sum_{i=1}^n K\left(\frac{x - X_i}{a_n}\right), \quad x \in \mathbb{R}^d \quad (1.2)$$

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where  $\{a_n, n \geq 1\}$  is a bandsequence (width of windows), that is, a sequence of positive numbers tending to 0 and satisfying

$$na_n^d \rightarrow +\infty \quad \text{as } n \rightarrow \infty. \quad (1.3)$$

As usual, we denote by  $\|g\|_p = (\int_{\mathbb{R}^d} |g(x)|^p dx)^{1/p}$ ,  $p \geq 1$ .

In [4] (see also [5]), Devroye proved that all types of  $L_1$ -consistency are equivalent to (1.3). The asymptotic normality of  $\{\|f_n - Ef_n\|_1, n \geq 1\}$  was studied by Csörgö and Horváth [1] and Horváth [7]. More recently, Giné, Mason and Zaitsev [13] considered the asymptotic normality of the  $L_1$ -norm density estimator process, Louani [17] and Lei, Wu and Xie [15] (see Lei and Wu [16] for density estimator in a Markov process) studied the large deviations in  $L_1(\mathbb{R}^d)$  for  $\{f_n(x), n \geq 1\}$ . For the uniform consistency, and uniform large deviations and uniform moderate deviations for  $\{f_n(x), n \geq 1\}$ , we refer to Einmahl and Mason [6], Giné and Guillou [10], Giné, Koltchinskii and Zinn [11], Louani [18], Gao [8] and the references therein. Giné and Mason [12] considered the law of the iterated logarithm for  $\{\|f_n - Ef_n\|_2^2 - E\|f_n - Ef_n\|_2^2, n \geq 1\}$  by the KMT approximation, and indicated that their methods do not extend to the cases  $\|\cdot\|_p$ ,  $p \neq 2$ . The purpose of this paper is to study the moderate deviations and the law of the iterated logarithm in  $L_1(\mathbb{R}^d)$  for  $\{f_n, n \geq 1\}$ . We find the best condition on the bandsequence such that  $\{\|f_n - Ef_n\|_1, n \geq 1\}$  satisfies the moderate deviation principle. A law of the iterated logarithm for  $\{\|f_n - Ef_n\|_1, n \geq 1\}$  is also obtained.

Let  $b_n, n \geq 1$  be a sequence of positive real numbers satisfying

$$\frac{n}{b_n} \rightarrow +\infty \quad \text{and} \quad \frac{n}{b_n^2} \rightarrow 0 \quad \text{as } n \rightarrow +\infty. \quad (1.4)$$

We introduce the following condition:

$$(H_1) \quad \int_{\mathbb{R}^d} (1 + |x|^{pd}) K^2(x) dx < \infty, \quad \int_{\mathbb{R}^d} |x|^{pd} f(x) dx < \infty, \quad \text{for some } p > 1.$$

**Remark 1.1.** If  $(H_1)$  holds, then

$$\int_{\mathbb{R}^d} \sqrt{f(x)} dx \leq \left( \int_{\mathbb{R}^d} (1 + |x|^{pd})^{-1} dx \right)^{1/2} \left( \int_{\mathbb{R}^d} (1 + |x|^{pd}) f(x) dx \right)^{1/2} < \infty,$$

and

$$\begin{aligned} \int_{\mathbb{R}^d} \sqrt{\int_{\mathbb{R}^d} \frac{1}{a_n^d} K^2\left(\frac{x-y}{a_n}\right) f(y) dy} dx &= \int_{\mathbb{R}^d} \sqrt{\int_{\mathbb{R}^d} K^2(z) f(x - a_n z) dz} dx \\ &\leq \left( \int_{\mathbb{R}^d} (1 + |x|^{pd})^{-1} dx \right)^{1/2} \left( \int_{\mathbb{R}^d} (1 + |x|^{pd}) \int_{\mathbb{R}^d} K^2(z) f(x - a_n z) dz dx \right)^{1/2} \\ &\leq \left( \int_{\mathbb{R}^d} (1 + |x|^{pd})^{-1} dx \right)^{1/2} \\ &\quad \times \left( \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} (1 + 2^{pd-1}(|u|^{pd} + |a_n v|^{pd})) K^2(u) f(v) du dv \right)^{1/2}, \end{aligned}$$

and so

$$\sup_{n \geq 1} \int_{\mathbb{R}^d} \sqrt{\int_{\mathbb{R}^d} \frac{1}{a_n^d} K^2\left(\frac{x-y}{a_n}\right) f(y) dy} dx < \infty.$$

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