

Asymptotic Normality of Kernel Density Estimates*

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Let X_1, \dots, X_n be a random samples taking values in \mathbf{R}^1 and having common density function $f(x)$. A class of estimators of $f(x)$ proposed by Rosenblatt^[1] has the form

$$f_n(x) = \frac{1}{nh_n} \sum_{i=1}^n K\left(\frac{x - X_i}{h_n}\right)$$

where K is a probability density function on \mathbf{R}^1 and $\{h_n\}$ is a sequence of positive numbers converging to zero. Call $f_n(x)$ the kernel estimator of $f(x)$. Many authors studied the large sample properties of $f_n(x)$ under the independent samples. For examples: Parzen^[2] and Sun^[3] showed the asymptotic normality of $(f_n(x) - Ef_n(x))/\sqrt{\text{Var} f_n(x)}$ respectively. Our paper aims to show the asymptotic normality of joint distribution of $\sqrt{nh_n}(f_n(x_1) - f(x_1), \dots, f_n(x_k) - f(x_k))$ under the condition that samples $\{X_n, n \geq 1\}$ is a sequence of stationary, ϕ -Mixing variables. Our main result is

Theorem Suppose that (I) $\sum_{n=1}^{\infty} \phi(n) < \infty$; (II) $(1 + |x|)K(x) < C < \infty$;

$\int |y|K(y)dy < \infty$; (III) $nh_n \rightarrow \infty$, $nh_n^3 \rightarrow 0$, as $n \rightarrow \infty$; (IV) $f(x)$ satisfies Lipschitz condition and joint density $g_i(y_1, y_2)$ of (X_1, X_i) is a continuous bounded function for any $i > 1$. Then $\sqrt{nh_n}(f_n(x_1) - f(x_1), \dots, f_n(x_k) - f(x_k))$ converges in distribution to Z for the distinct point (x_1, \dots, x_k) where $f(x_i) > 0$ ($i = 1, \dots, k$). Where Z is a k -dimensional normal vector and having mean vector 0 and diagonal covariance matrix $B = (b_{ij})$, $b_{ii} = f(x_i) \int K^2(y)dy$, $i = 1, \dots, k$.

References

- [1] Rosenblatt, M., Ann. Math. Statist., 27(1956), 832—837.
- [2] Parzen, E., AMS, 33(1962), 1065—1076.
- [3] Sun Zhigang, Acta. Math. Sinica, 27(1984), 769—782.

* Received Mar. 18, 1988. Revised to this note May 10, 1990.