

Prediction of nonparametric regression models with long memory data.

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abstract: In this work, the kernel estimator of regression function is investigated for stationary long memory (long range dependence) of the proposed predictor. Under general conditions , we establish the asymptotic normality and propose a confidence bands of these estimators .

Keyword : *regression function, Kernel estimation, Asymptotic normality ,prediction, Long memory,*

Related publications:

- Giraitis, L Koul, H L Surgailis, D. Asymptotic normality of regression estimators with long memory errors, *Statist Probab Lett*, 29: 317-335 (1996),.
- Palma. W. Long-Memory Time Series, Theory and Methods,, John Wiley, (2007).
- Wang. L, Wang. M, Asymptotics of estimators for nonparametric multivariate regression models with long memory, *Appl. Math. J. Chinese Univ.* 34(4): 403-422 (2019),



Introduction & motivation

One of the most important problems in time series analysis is prediction of future observations.

Namely, given the observed series Z_1, Z_2, \dots, Z_n , the aim is to predict the unobserved value Z_{n+p} , for some integer $p \geq 1$. The classical approach to this problem is to find some nonparametric estimate of the regression function defined by $m(x) = E(Y|X=x)$ where $(X, Y) \in \mathbb{R}^p \times \mathbb{R}$.

Kernel regression estimators have been widely studied in the literature when the data are (iid, mixing and association condition). Nadaraya and Watson (1964), Schuster (1972), Devroye and Györfi (1985), Bosq (1998), Cai (2001), Roussas (2000) for associated data, Laib and al. (2010, 2011) for ergodic data, Ferraty and al (2002, 2007) for functional data and Wang and al (2019) when the error is long memory.

Long-memory (LM) or Long range dependence (LRD) is a phenomenon that is associated with strong correlations between the present and past values of a stochastic process. In recent years, LM has become an important tool for analyzing dependent data. They have played roles in various areas (physical sciences, finance, hydrology and climatology ...) see, for example, Beran (2013), Doukhan and al. (2003), and Giraitis and al. (2012) and references therein.



Methods & Model

Definition :

for a stationary process (X_t) , $t \in Z$, is said long-memory if the auto-covariance function

$\gamma(k) = \text{cov}(X_{t+k}, X_t)$ is not summable, i.e. $\sum_{k=1}^{\infty} \gamma(k) = +\infty$.

Model

Let Z_1, Z_2, \dots, Z_n be real-valued with long memory random variables forming a strictly stationary sequence, and let φ be a real-valued function defined on R . The prediction problem in a discrete parameter time series given by

$$X_i = (Z_i, \dots, Z_{i+d-1}), Y_i = \varphi(Z_{i+d}), i \geq 1, d \geq 1 \quad \gamma_Z(i) = \text{cov}(Z_0, Z_i) = 0$$

and the regression function

$$m(x) = \mathbb{E}[\varphi(Y_1) | X_1 = x] = \mathbb{E}[\varphi(Z_{d+1}) | Z_1, \dots, Z_d], x \in \mathbb{R}^d$$

The quantity $\varphi(Z_{j+d})$ is to be predicted on the basis of the past d r.v.'s $Z_1, Z_2, \dots, Z_{j+d-1}$. The proposed kernel estimator (Nadaraya-Watson) given by

$$m_n(x) = \frac{g_n(x)}{f_n(x)} \quad \text{where}$$

$$f_n(x) = \frac{1}{nh^d} \sum_{j=1}^n K\left(\frac{x - X_j}{h}\right), g_n(x) = \frac{1}{nh^d} \sum_{j=1}^n \varphi(Y_j) K\left(\frac{x - X_j}{h}\right)$$



Results and Conclusions

- The natural estimate $m_n^*(x)$ for $m(x)$ is defined by the conditional p.d.f (noted $w_n(\cdot/x)$) of Y_1 , given $X_1 = x$. where

$$m_n^*(x) = \int_{\mathbb{R}} \varphi(y)w_n(dy | x),$$

- It is shown, however that $m_n^*(x) - m_n(x) \xrightarrow[n \rightarrow \infty]{} 0, x \in \mathbb{R}$, pointwise,

- The result of this work is proving asymptotic normality of our kernel estimator.

$$\sqrt{nh^d} [m_n(x) - m(x)] \xrightarrow[n \rightarrow \infty]{d} N(0, \sigma^2(x)), x \in \mathbb{R}$$

where

$$\sigma^2(x) = \sigma_{\varphi}^2(x) \int_{\mathbb{R}^d} \frac{K^2(t)dt}{f(x)}, (f(x) > 0),$$

$$\text{and } \sigma_{\varphi}^2(x) = \mathcal{E} \{ [\varphi(Y_1) - m(x)]^2 | X_1 = x \}$$

Conclusion : We study in this work the kernel estimator of regression function and solve the prediction problem in a discrete parameter time series under long memory dependence . And we can derive the confidence interval.

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4th Conference on
**Statistics and
Data Science**
Salvador, Brazil (online)
December 1-3, 2022