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STRONG CONSISTENCY OF A SEQUENTIAL ESTIMATOR OF A PROBABILITY DENSITY FUNCTION

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1. Introduction.

Consider a sequence $X_1, X_2, X_3 \cdots$ of independent identically distributed m-dimensional random vectors having a probability density function, f(x). Van Ryzin (1969) has shown that under appropriate conditions, estimates of the form

(1)
$$f_n(x) = n^{-1} \sum_{j=1}^n K_n(x, X_j),$$

where

$$K_n(x, X_i) = h_n^{-m} K(h_n^{-1}(x - X_i))$$

are strongly consistent. That is, if x is contained in the continuity set C(f), then $f_n(x) \to f(x)$ with probability one. Here $K(u) = K(u_1, u_2, \dots, u_m)$ is a real valued, Borel measurable function on R^m , where R^m is the m-fold convolution of the real line, such that

(2)
$$K(u)$$
 is a density on R^m

$$\sup_{u \in R^m} |K(u)| < \infty$$

(4)
$$||u||K(u) \to 0$$
 as $||u||^2 = \sum_{i=1}^m u_i^2 \to \infty$

and $\{h_n\}$ is a sequence of numbers such that

(5)
$$h_n > 0, \quad n = 1, 2, \cdots; \quad \lim_{n \to \infty} h_n = 0, \quad \text{and} \quad \lim_{n \to \infty} nh_n = \infty.$$

Yamato, (1970) considers a modified estimator of the form

(6)
$$\hat{f}(x) = n^{-1} \sum_{j=1}^{n} h_j^{-m} K(h_j^{-m}(x - X_j))$$

and shows the weak consistency of this estimator. (That is $\lim_{n\to\infty} E|\hat{f}_n(x)-f(x)|^2=0$ at all points x). It is our purpose to show, using Van Ryzin's method that this modified estimator is also strongly consistent. However, because of the special form that this estimator takes, Van Ryzin's conditions can be somewhat weakened.

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2. Strong Consistency.

Pointwise: In order to show strong consistency we will use the following lemma which is proved in Van Ryzin (1969).

LEMMA 1. Let $\{Y_n\}$ and $\{Y'_n\}$ be two sequences of random variables on a probability space (Ω, F, P) . Let $\{F_n\}$ be a sequence of Borel fields, $F_n \subset F_{n+1} \subset F$, where Y_n and Y'_n are measurable with respect to F_n . If

- (i) $0 \leq Y_n$ a.e.
- (ii) $EY_1 < \infty$
- (iii) $E[Y_{n+1}|F_n] \leq Y_n + Y'_n$ a. e.
- (iv) $\sum_{n=1}^{\infty} E|Y_n'| < \infty$

then Y_n converges a.e. to a finite limit.

THEOREM 1. If K(u) satisfies (2)—(4) and $\{h_n\}$ is a monotone decreasing sequence of numbers satisfying (5) and if in addition

$$\sum_{n=1}^{\infty} n^{-2} h_n^{-m} < \infty$$

(8)
$$\lim_{n \to \infty} n^{-1} \sum_{j=1}^{n} h_j^{-m} h_n^m = \alpha , \qquad 0 \le \alpha \le 1$$

then $\hat{f}_n(x) \rightarrow f(x)$ with probability one if $x \in C(f)$.

PROOF. $E\hat{f}_n(x) \to f(x)$ as $n \to \infty$ (Yamato (1970)). Thus, it is sufficient to show $\hat{f}_n(x) - E\hat{f}_n(x) \to 0$ with probability one as $n \to \infty$.

Let $Y_n = \{\hat{f}_n(x) - E\tilde{f}_n(x)\}^2$ and $F_n =$ Borel field generated by X_1, X_2, \dots, X_n and note that $\lim E(Y_n) = 0$ (Yamato (1970)). Since,

(9)
$$\hat{f}_n(x) = n^{-1}(n-1)\hat{f}_{n-1}(x) + n^{-1}h_n^{-m}K(h_n^{-1}(x-X_n)),$$

it follows that

$$\begin{split} \hat{f}_{n+1}(x) - E\hat{f}_{n+1}(x) &= \hat{f}_{n}(x) - E\hat{f}_{n}(x) - (n+1)^{-1} \{ \hat{f}_{n}(x) - E\hat{f}_{n}(x) \} \\ &+ (n+1)^{-1} \{ h_{n+1}^{-m} K(h_{n+1}^{-1}(x-X_{n+1})) - h_{n+1}^{-m} E(K(h_{n+1}^{-1}(x-X_{n+1}))) \} \; . \end{split}$$

Thus.

$$E[Y_{n+1}|F_n] = Y_n + (n+1)^{-2}Y_n - 2(n+1)^{-1}Y_n + (n+1)^{-2}h_{n+1}^{-2m} \operatorname{Var} K(h_{n+1}^{-1}(x-X))$$
.

Let $Y'_n = (n+1)^{-2} Y_n - 2(n+1)^{-1} Y_n + (n+1)^{-2} h_{n+1}^{-2m} \text{ Var } K(h_{n+1}^{-1}(x-X)).$

In order to use Lemma 1 we must verify

(10)
$$\sum_{n=1}^{\infty} E|Y'_n| < \infty.$$

But,

$$\sum_{n=1}^{\infty} (n+1)^{-1} E|Y_n| = \sum_{n=1}^{\infty} (n+1)^{-1} \operatorname{Var} \hat{f}_n(x).$$

Now, Yamato (1970), using a monotone sequence satisfying (5) and (8) has shown (Theorem 3) that

(11)
$$\lim_{n\to\infty} nh_n^m \operatorname{Var} \hat{f}_n(x) = \alpha f(x) \int_{\mathbb{R}^m} K^2(y) dy.$$

Thus $n^{-1} \operatorname{Var} \hat{f}_n(x) = 0 (n^{-2} h_n^{-m})$ for large n. Hence, $\sum_{n=1}^{\infty} (n+1)^{-1} \operatorname{Var} \hat{f}_n(x) < \infty$ using condition (7).

The fact that $\sum_{n=1}^{\infty} (n+1)^{-2} h_{n+1}^{-2m} \operatorname{Var} K(h_{n+1}^{-1}(x-X)) < \infty$ is shown to follow from condition (7) by Van Ryzin (1969).

Thus $\sum_{n=1}^{\infty} E|Y'_n| < \infty$ and the result follows by using Lemma 1 and the fact that the mean square limit and the almost sure limit coincide with probability one.

Uniform: Let $k(t) = k(t_1, \dots, t_m) = \int e^{it'u} K(u) du$ where $t'u = \sum_{j=1}^m t_j u_j$. Also let $\phi_n(t) = n^{-1} \sum_{j=1}^n e^{it'Xj}$ and let $\phi(t) = Ee^{it'X}$.

THEOREM 2. If K(u) satisfies conditions (2)—(4) and $\{h_n\}$ is a monotone decreasing sequence of real numbers satisfying (5) and if both $h_n h_{n+1}^{-1} \to 1$ and $n h_n^{-2m} \to \infty$ as $n \to \infty$, and if

$$(10) \qquad \qquad \sum_{n=1}^{\infty} (nh_n^m)^{-2} < \infty$$

$$\sum_{n=1}^{\infty} \frac{1}{n h_n^{2m-1}} \left| \frac{1}{h_{n+1}} - \frac{1}{h_n} \right| < \infty$$

and if also

$$\int |k(t)| dt < \infty$$

where |k(u)| is non-decreasing on u < 0 and non-increasing on $u \ge 0$, then if f(x) is uniformly continuous on R^m , $\sup |\hat{f}_n(x) - f(x)| \to 0$ with probability one as $n \to \infty$.

PROOF. (All integrals are over R^m , unless otherwise specified.) We first prove $\sup_x |\hat{f}_n(x)E\hat{f}_n(x)| \to 0$ a.s. as $n \to \infty$. Since both k(u) and K(u) are in L_1 , using the inversion theorem for a Fourier transform

$$\begin{split} \sup_{x} |\hat{f}_{n}(x) - E\hat{f}_{n}(x)| &= \sup_{x} \left| (2\pi)^{-1} \int n^{-1} \sum_{j=1}^{n} \left[e^{iu \cdot X_{j}} - \phi(u) \right] k(h_{j}u) e^{-iu \cdot x} du \right| \\ & \leq (2\pi)^{-m} \int \left| n^{-1} \sum_{j=1}^{n} \left[e^{iu \cdot X_{j}} - \phi(u) \right] k(h_{j}u) \right| du \\ &= (2\pi)^{-m} \int \left| n^{-1} \sum_{j=1}^{n} \left[e^{iu \cdot X_{j}} - \phi(u) \right] \frac{k(h_{j}u)}{k(h_{n}u)} \right| |k(h_{n}u)| du \;. \end{split}$$

Using Schwartz's inequality, this is less than

$$\begin{split} &(2\pi)^{-m} \Big\{ \int \Big| k(h_n u) \, | \, du \Big\}^{\frac{1}{2}} \Big\{ \int \Big| n^{-1} \sum_{j=1}^n \big[e^{iu \cdot X_j} - \phi(u) \big] \, \frac{k(h_j u)}{k(h_n u)} \Big|^2 \, | \, k(h_n u) \, | \, du \Big\} \\ &= (2\pi)^{-m} \Big\{ \int \Big| h_n^{-m} \, | \, k(u) \, | \, du \Big\}^{\frac{1}{2}} \Big\{ \int \Big| n^{-1} \sum_{j=1}^n \big[e^{iu \cdot X_j} - \phi(u) \big] \, \frac{k(h_j n)}{k(h_n u)} \Big|^2 \, | \, k(h_n u) \, | \, du \Big\} \, . \end{split}$$

Since $\int |\mathit{k}(\mathit{u})| \, d\mathit{u} < \infty$ by assumption (12), we need only consider the term

$$\int h_n^{-m} |n^{-1} \sum_{j=1}^n \left[e^{iu \cdot X_j} - \phi(u) \right] - \frac{k(h_j u)}{k(h_n u)} \bigg|^2 |k(h_n u)| du$$

which we will denote by Y_n . Taking expectations we get

$$EY_n = \int h_{n_1}^{-m} E |n^{-1} \sum_{i=1}^n \left[e^{iu \cdot X_j} - \phi(u) \right] \frac{k(h_j u)}{k(h_n u)} \Big|^2 |k(h_n u)| du.$$

But,

$$E|e^{iuX_j}-\phi(u)|^2 = E\lceil e^{iuX_j}-\phi(u)\rceil\lceil e^{-iuX_j}-\overline{\phi(u)}\rceil = 1-|\phi(u)|^2$$

so that, we obtain

(13)
$$EY_n = n^{-2} h_n^{-m} \int (1 - |\phi(u)|^2) \sum_{j=1}^n \left| \frac{k(h_j u)}{k(h_n u)} \right|^2 |k(h_n u)| du .$$

However, since |k(u)| is non-decreasing on u < 0 and non-increasing on $u \ge 0$, we have

(14)
$$\left| \frac{k(h_j u)}{k(h_n u)} \right| \leq 1, \quad j \leq n.$$

Thus, using inequality (14) in equation (13) we get

$$EY_n \le n^{-1}h_n^{-m} \int |k(h_n u)du| = n^{-1}h_n^{-2m} \int |k(u)du$$
.

Condition (12) and the assumption $nh_n^{2m} \to \infty$ ensure that EY_n tends to zero as $n \to \infty$. We now prove Y_n converges with probability one. Write,

$$\begin{split} \boldsymbol{Y}_{n+1} &= h_{n+1}^{-m} \int (n+1)^{-2} \left| \sum_{j=1}^{n+1} \left \lfloor e^{iu \cdot \boldsymbol{X} j} - \phi(u) \right \rfloor \frac{k(h_{j}u)}{k(h_{n+1}u)} \right|^{2} |k(h_{n+1}u)| \, du \\ &= (n+1)^{-2} h_{n+1}^{-m} \int \left| \sum_{j=1}^{n} \left \lfloor e^{iu \cdot \boldsymbol{X} j} - \phi(u) \right \rfloor \frac{k(h_{j}u)}{k(h_{n}u)} + e^{iu \cdot \boldsymbol{X} n + 1} - \phi(u) \right|^{2} |k(h_{n+1}u)| \, du \; . \end{split}$$

Let F_n be defined as in Theorem 1. We now expand the first term in the integral using the fact that something of the form $|z|^2 = z\overline{z}$ and that

$$E(e^{iu\cdot X_{n+1}}-\phi(u)|F_n)=0$$
 and $E(e^{-iu\cdot X_{n+1}}-\overline{\phi(u)}|F_n)=0$.

Thus,

$$\begin{split} E \left[\left. Y_{n+1} \right| F_n \right] &= (n+1)^{-2} h_{n+1}^{-m} \int \left| \sum_{j=1}^n \left[e^{iu'X_j} - \phi(u) \right] \frac{k(h_j u)}{k(h_n u)} \right|^2 |k(h_{n+1} u)| \, du \\ &+ h_{n+1}^{-m} (n+1)^{-2} \int (1 - |\phi(u)|^2) |k(h_{n+1} u)| \, du \; . \end{split}$$

But $(1-|\phi(u)|^2) \le 1$ and $|k(u)| \le 1$ since we assume K(u) is a density function. Thus,

$$\begin{split} E [Y_{n+1}|F_n] & \leq n^{-2} h_{n+1}^{-m} \int \left| \sum_{j=1}^n \left[e^{iu'X_j} - \phi(u) \right] \frac{k(h_j u)}{k(h_n u)} \right|^2 |k(h_n u)| \, du \\ & + n^{-2} h_{n+1}^{-2m} \int |k(u)| \, du \; . \end{split}$$

Adding and subtracting Y_n we obtain

Let,

(15)
$$V_n = (nh_{n+1}^{2m})^{-1} \int |k(u)| du$$

and

(16)
$$U_n = n^{-2} (h_n^{-m} - h_{n+1}^{-m}) \int \left| \sum_{j=1}^n \left[e^{iu \cdot X_j} - \phi(u) \right] \frac{k(h_j u)}{k(h_n u)} \right|^2 |k(h_n u)| du.$$

Note that V_n is not a random variable. Hence $EV_n = V_n$, and, using condition (10), we have $\sum_{n=1}^{\infty} E|V_n| = \sum_{n=1}^{\infty} |V_n| < \infty$. Thus, in order to use Lemma 1 we need to show now that $\sum_{n=1}^{\infty} E|U_n| < \infty$. From (16) we have, taking expectations,

(17)
$$E|U_{n}| = n^{-2} |h_{n}^{-m} - h_{n+1}^{-m}| \int (1 - |\phi(u)|^{2}) \sum_{j=1}^{n} \left| \frac{k(h_{j}u)}{(kh_{n}u)} \right|^{2} |k(h_{n}u)| du$$

$$\leq n^{-2} |h_{n}^{-m} - h_{n+1}^{-m}|h_{n}^{-m} \int n |k(u)| du .$$

By using the fact that $(1-t^m)=\Big(\sum\limits_{j=1}^mt^{j-1}\Big)(1-t)$ and $h_nh_{n+1}^{-1}\to 1$, Van Ryzin (1969) shows

$$|h_{n+1}^{-m}-h_n^{-m}| \sim mh_n^{m-1}|h_{n+1}^{-1}-h_n^{-1}|$$
 .

Thus the upper bound in (17) is asymptotically equivalent to $\frac{m}{nh_n^{2m-1}}|h_{n+1}^{-1}-h_n^{-1}|\times\int |k(u)|\,du$. Hence, using condition (11), $\sum_{n=1}^{\infty}E|U_n|<\infty$. Thus conditions (iii) and (iv) of Lemma 1 are satisfied, where $Y_n'=U_n+V_n$.

The conditions of Lemma 1 are satisfied for Y_n and so Y_n tends to zero almost everywhere. Thus,

$$\sup_{x} |\hat{f}_{n+1}(x) - E\hat{f}_{n+1}(x)| \to 0 \quad \text{a. s.} \quad \text{as} \quad n \to \infty.$$

Now,

$$\begin{split} \sup_{x} |E\hat{f}_{n}(x) - f(x)| &= \sup_{x} \left| \int \left(n^{-2} \sum_{j=1}^{n} K_{j}(u) f(x-u) - n^{-1} \sum_{j=1}^{n} K_{j}(u) f(x) \right) du \right| \\ & \leq \sup_{x} \int n^{-1} \left| \sum_{j=1}^{n} K_{j}(u) \right| |f(x-u) - f(x)| \, du \\ & \leq \sup_{x} \int_{\|u\| \leq \delta} n^{-1} \left| \sum_{j=1}^{n} K_{j}(u) \right| |f(x-u) - f(x)| \, du \\ & + \sup_{x} \int_{\|u\| \leq \delta} |f(x-u) - f(x)| \, n^{-1} \sum_{j=1}^{n} K_{j}(u) du \\ & \leq \sup_{x} \sup_{\|u\| \leq \delta} |f(x-u) - f(x)| + 2 \sup_{x} n^{-1} f(x) \sum_{j=1}^{n} \int_{\|u\| > \delta/hj} K(u) du \end{split}$$

Since f(x) is uniformly continuous, the first term can be made arbitrarily small by choosing δ sufficiently small. In the second term, notice that for this fixed δ

$$f(x) \int_{\|u\| \ge \delta/h_n} K(u) du \to 0$$

as $n \to \infty$ at all points x. Hence the Cesàro sum approaches zero and the proof is complete.

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